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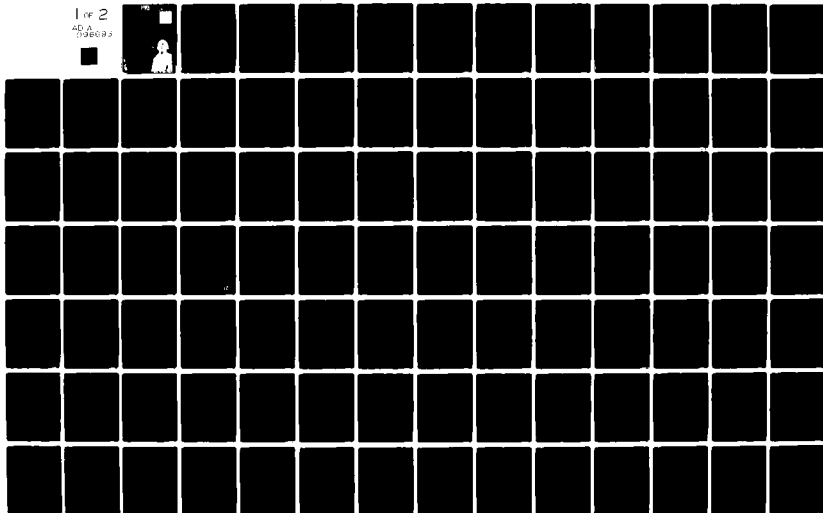
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A USER'S MANUAL FOR SENSUMT:

A Penalty Function Computer Program for Solution,  
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Calculation in Parametric Nonlinear Programs.

10 by 7 Scientific Dept.  
Anthony V. Fiacco  
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output is included. For completeness, the computer listing and a brief description of all the subroutines comprising SENSUMT, many of which are taken intact from a previously developed program SUMT-Version 4, are also provided.

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This manual is intended to serve as a guide for coding, solving, and conducting sensitivity and optimal value bound analysis for parametric nonlinear programming problems with the computer programming model SENSUMT. The basic sensitivity results and bound calculation techniques are briefly reviewed and the algorithms implementing them are presented. A procedure for coding problems for SENSUMT and detailed illustrations of the computer output is included. For completeness, the computer listing and a brief description of all the subroutines comprising SENSUMT, many of which are taken intact from a previously developed program SUMT-Version 4, are also provided.

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This latest version of the SENSUMT program is an experimental program that is still evolving, though it has been used routinely at The George Washington University over the past several years to solve intermediate sized (up to about 100 variables and constraints) NLP problems and to conduct solution sensitivity analysis. A major part of it is the product of many theoretical, computational and software developments contributed by others. Several subroutines have been taken essentially intact from existing codes, most notably from SUMT-Version 4, coded by W. C. Mylander, R. L. Holmes, and G. P. McCormick; the first version of SENSUMT coded by R. L. Armacost and W. C. Mylander; and an extended version of SENSUMT coded subsequently by Armacost. The authors wish especially to acknowledge these contributions, along with the first program to implement some of the techniques used by this general approach for the calculation of sensitivity analysis, coded by B. Causey.

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1. Introduction

This is a manual intended to facilitate the use of the computer program SENSUMT. This program is capable of solving general parametric nonlinear programs and conducting a sensitivity analysis using the Sequential Unconstrained Minimization Technique [10]. It has recently been modified to include the calculation of piecewise linear parametric bounds on the optimal value function of classes of nonlinear programming problems for which this function is convex or concave. In addition, it can calculate similar bounds on a class of separable non-convex right-hand side programs, once appropriate convex overestimating and underestimating programs are formulated.

SENSUMT is the outgrowth of a routine [3] which first implemented a penalty function sensitivity approach. This was motivated by results given by Fiacco and McCormick [10] for a particular class of perturbations. The algorithmic procedure for this routine was suggested by Fiacco. It was later refined and recoded by Mylander [12] making use of

several subroutines from the SUMT-Version 4 computer program coded by Mylander, Holmes, and McCormick [13]. These routines were subsequently completely integrated with SUMT-Version 4 by Armacost and Mylander [7]. Armacost [1] further revised the program to implement the generalized version of the sensitivity theory developed by Fiacco [9] for general parameter perturbations. This routine, again following a procedure suggested by Fiacco, was recently modified by Ghaemi to conduct piecewise linear parametric bound calculation on the optimal value function of certain classes of nonlinear programming problems subject to large parametric changes [11].

This latter version of the model, designated "SENSUMT," is compiled at the Center for Academic and Administrative Computing of The George Washington University.

The various sections of this manual are arranged as follows. Section 2 reviews the basic sensitivity results and presents the steps of the algorithm implementing sensitivity calculation via a sequential unconstrained minimization algorithm.

Section 3 briefly reviews the procedure for calculating piecewise linear parametric bounds on the optimal value function of certain classes of parametric nonlinear programming problems. It presents the relevant algorithms that have been implemented.

Section 4 gives the details regarding the procedure for coding the problems for solution, sensitivity analysis, and bound calculation with SENSUMT.

Section 5 provides the codes and corresponding computer solution for two illustrative examples that are taken from [11]. For clarity, an annotated computer listing of the input and output for the first example is also provided.

Section 6 gives a description and listing of the various subroutines comprising SENSUMT. This section is included to make the

manual self-contained. With the exception of the subroutines BOUND, PERT, and TRANS, and the new version of program MAIN, the material in this section with a few minor modifications has been taken from [1] and [13].

## 2. Sensitivity Analysis in Nonlinear Programming

### 2.1 Basic Sensitivity Results

The parametric mathematical programming problem considered by Fiacco [9] is of the following general form:

$$\begin{aligned} & \underset{x \in E^n}{\text{minimize}} && f(x, \epsilon) \\ & \text{subject to} && g_i(x, \epsilon) \geq 0, \quad i=1, \dots, m, \\ & && h_j(x, \epsilon) = 0, \quad j=1, \dots, p, \end{aligned} \quad P(\epsilon)$$

where  $x$  is the usual vector of variables and  $\epsilon$  is a  $k$ -component vector of numbers called "parameters." It is desired to analyze the behavior of a solution vector  $x(\epsilon)$  and the optimal solution value  $f^*(\epsilon) \equiv f[x(\epsilon), \epsilon]$  near some given value of  $\epsilon$ . Without loss of generality, assume that the parameter vector of interest is  $\epsilon=0$ .

The Lagrangian for Problem  $P(\epsilon)$  is defined as

$$L(x, u, w, \epsilon) \equiv f(x, \epsilon) - \sum_{i=1}^m u_i g_i(x, \epsilon) + \sum_{j=1}^p w_j h_j(x, \epsilon). \quad (2.1)$$

The sensitivity results are based on the following four assumptions:

- A1 - The functions defining Problem  $P(\epsilon)$  are twice continuously differentiable in  $(x, \epsilon)$  in a neighborhood of  $(x^*, 0)$ .
- A2 - The second order sufficient conditions for a local minimum of Problem  $P(0)$  hold at  $x^*$  with associated Lagrange multipliers  $y^*$  and  $w^*$ .

A3 — The gradients  $\nabla_x g_i(x^*, 0)$ , for all  $i$  such that  $g_i(x^*, 0) = 0$ , and  $\nabla_x h_j(x^*, 0)$ ,  $j=1, \dots, p$  are linearly independent.

A4 — Strict complementary slackness holds at  $x^*$  when  $\epsilon=0$  [i.e.,  $u_i^* > 0$  for all  $i$  such that  $g_i(x^*, 0) = 0$ ].

Under the above assumptions, Fiacco [9] established the following generalization of Theorem 6 in [10].

Lemma 2.1 [local characterization of a Kuhn-Tucker triple]: If assumptions A1, A2, A3, and A4 hold for Problem  $F(\epsilon)$  at  $(x^*, 0)$ , then

- (a)  $x^*$  is a local isolated minimizing point of Problem  $P(0)$  and the associated Lagrange multipliers  $u^*$  and  $w^*$  are unique;
- (b) for  $\epsilon$  in a neighborhood of 0, there exists a unique once continuously differentiable vector function  $y(\epsilon) = (x(\epsilon), u(\epsilon), w(\epsilon))^T$  satisfying the second order sufficient conditions for a local minimum of Problem  $P(\epsilon)$  such that  $y(0) = (x^*, u^*, w^*)^T = y^*$ , and hence,  $x(\epsilon)$  is a locally unique, local minimum of Problem  $P(\epsilon)$  with associated unique Lagrange multipliers  $u(\epsilon)$  and  $w(\epsilon)$ ; and
- (c) for  $\epsilon$  near 0, the set of binding inequalities is unchanged, strict complementary slackness continues to hold, and the binding constraint gradients are linearly independent at  $x(\epsilon)$ .
- (d) (Armstrong and Fiacco [3]), for  $\epsilon$  near 0, the gradient of the optimal value function is

$$\nabla_{\epsilon} f^*(\epsilon) = \nabla_{\epsilon} L(y(\epsilon), \epsilon), \quad (2.2)$$

- (e) which also means that, for  $\epsilon$  near 0, the Hessian of the optimal value function is

$$\nabla_{\epsilon}^2 f^*(\epsilon) = \nabla_{\epsilon}^2 L(y(\epsilon), \epsilon). \quad (2.3)$$



The above results provide a characterization of a local solution of Problem  $P(\epsilon)$  and its associated optimal Lagrange multipliers near  $\epsilon=0$ . They show that the Kuhn-Tucker triple  $y(\epsilon)$  is unique and well behaved, under the given conditions. Since  $y(\epsilon)$  is once differentiable, the partial derivatives of the components of  $y(\epsilon)$  are well defined. This fact and assumption A1 also mean that the functions defining Problem  $P(\epsilon)$  are once continuously differentiable functions of  $\epsilon$  along the "solution trajectory"  $x(\epsilon)$  near  $\epsilon=0$ , and the Lagrangian is a once continuously differentiable function of  $\epsilon$  along the "Kuhn-Tucker point trajectory." The above results constitute the structure for numerous developments and extensions, many of which have been established by Fiacco [9] and Armacost and Fiacco [2 - 6].

The realization of this theorem for the parametric right hand side problem of special interest in the present study is treated in detail by Armacost and Fiacco [4]. The parametric right hand side problem is the following important realization of  $P(\epsilon)$ :

$$\begin{aligned} &\text{minimize } f(x) \\ &\text{subject to } g_i(x) \geq \epsilon_i, \quad i=1, \dots, m, \\ &\quad h_j(x) = \epsilon_{j+m}, \quad j=1, \dots, p. \end{aligned} \quad R(\epsilon)$$

The Lagrangian for  $R(\epsilon)$  is

$$L(x, u, w) \equiv f(x) - \sum_{i=1}^m u_i (g_i(x) - \epsilon_i) + \sum_{j=1}^p w_j (h_j(x) - \epsilon_{j+m}).$$

As is evident from the Lagrangian, the results (d) and (e) of Lemma 2.1 for Problem  $R(\epsilon)$  simplify respectively to

$$(d') \quad \nabla_{\epsilon} f^*(\epsilon) = \begin{bmatrix} u(\epsilon) \\ -w(\epsilon) \end{bmatrix}^T \quad (2.4)$$

and

$$(e') \quad \nabla_{\epsilon}^2 f^*(\epsilon) = \begin{bmatrix} \nabla_{\epsilon} u(\epsilon) \\ -\nabla_{\epsilon} w(\epsilon) \end{bmatrix}. \quad (2.5)$$

Fiacco [9] has shown that the class of algorithms based on twice continuously differentiable penalty functions (specifically, using the logarithmic-quadratic loss penalty function) can be used to estimate  $y(\epsilon)$  and its derivatives in a neighborhood of  $\epsilon=0$ , for the general problem  $P(\epsilon)$ . Minimization of the penalty function with penalty parameter  $r$  yields a solution of a perturbation of the Kuhn-Tucker system in a neighborhood of  $(\epsilon, r) = (0, 0)$ . Armacost and Fiacco [3] define an optimal value penalty function and obtain first- and second-order sensitivity estimates which converge to the corresponding sensitivities for the optimal value function for Problem  $P(\epsilon)$ .

The logarithmic-quadratic penalty function is

$$W(x, \epsilon, r) \equiv f(x, \epsilon) - r \sum_{i=1}^m \ln g_i(x, \epsilon) + (1/2r) \sum_{j=1}^p h_j^2(x, \epsilon). \quad (2.6)$$

Lemma 2.2 (Fiacco [9, Theorem 3.1]): If the assumptions A1 through A4 hold, then in a neighborhood of  $(\epsilon, r) = (0, 0)$  there exists a unique once continuously differentiable vector function  $y(\epsilon, r) = [x(\epsilon, r), u(\epsilon, r), w(\epsilon, r)]^T$  satisfying

$$\begin{aligned} \nabla_x L(x, u, w, \epsilon) &= 0, \\ u_i g_i(x, \epsilon) &= r, \quad i=1, \dots, m, \\ h_j(x, \epsilon) &= w_j r, \quad j=1, \dots, p, \end{aligned} \quad (2.7)$$

with  $y(0, 0) = (x^*, u^*, w^*)$ , and such that for any  $(\epsilon, r)$  near  $(0, 0)$  and  $r > 0$ ,  $x(\epsilon, r)$  is a locally unique unconstrained local minimizing point of  $W(x, \epsilon, r)$ , with  $g_i[x(\epsilon, r), \epsilon] > 0$ ,  $i=1, \dots, m$ , and  $\nabla_x^2 W[x(\epsilon, r), \epsilon, r]$  positive definite.

The relevance of Equations (2.7) is the fact that, under the given conditions, when  $r=0$ , they are necessary conditions that must hold at a local solution of  $P(0)$  and, with  $r > 0$ , they are necessary conditions for an unconstrained minimum of  $W(x, \epsilon, r)$ . The latter fact can be made obvious by solving for  $u_i$  and  $w_j$  in (2.7) and obtaining

$$\begin{aligned}
\nabla_x L(x, u, w, \varepsilon) &= \nabla_x f - \sum_i u_i \nabla_x g_i + \sum_j w_j \nabla_x h_j \\
&= \nabla_x f - r \sum_i (1/g_i) \nabla_x g_i + (1/r) \sum_j h_j \nabla_x h_j \\
&= \nabla_x W(x, \varepsilon, r) .
\end{aligned} \tag{2.8}$$

Thus, if  $y(\varepsilon, r)$  is a solution of (2.7), then

$$\nabla_x W[x(\varepsilon, r), \varepsilon, r] = \nabla_x L[x(\varepsilon, r), u(\varepsilon, r), w(\varepsilon, r), ] = 0 . \tag{2.9}$$

This explicit connection between the optimality conditions of local solutions of  $P(\varepsilon)$  and unconstrained minima of  $W(x, \varepsilon, r)$  makes it possible to approximate information characterizing a local solution of  $P(\varepsilon)$  by algorithmic calculations associated with utilizing  $W(x, \varepsilon, r)$  to solve  $P(\varepsilon)$ . In particular, differentiating (2.9) with respect to  $\varepsilon$  yields

$$\nabla_x^2 W \nabla_\varepsilon x + \nabla_{\varepsilon x}^2 W = 0 ,$$

and using the fact that  $\nabla_x^2 W$  is positive definite (a conclusion of Lemma 2.2) yields

$$\nabla_\varepsilon x = -\nabla_x^2 W^{-1} \nabla_{\varepsilon x}^2 W , \tag{2.10}$$

evaluated, of course, at  $x(\varepsilon, r)$ . Given  $\nabla_\varepsilon x(\varepsilon, r)$ , the derivatives of the multipliers,  $\nabla_\varepsilon u_i(\varepsilon, r)$  and  $\nabla_\varepsilon w_j(\varepsilon, r)$ , can then be calculated by differentiating the last two systems of equations of (2.7) at  $x(\varepsilon, r)$  with respect to  $\varepsilon$ .

Lemma 2.3 (Fiacco [9]): For  $\varepsilon$  in a neighborhood of  $\varepsilon=0$ , it follows that

- (a)  $\lim_{r \rightarrow 0^+} y(\varepsilon, r) = y(\varepsilon, 0) = y(\varepsilon)$ , the Kuhn-Tucker triple characterized in conclusion (b) of Lemma 2.1; and
- (b)  $\lim_{r \rightarrow 0^+} \nabla_\varepsilon y(\varepsilon, r) = \nabla_\varepsilon y(\varepsilon, 0) = \nabla_\varepsilon y(\varepsilon)$ .

Under the conditions of Lemma 2.1,  $x(\epsilon, r)$  is a locally unique minimizing point of  $W(x, \epsilon, r)$ . Define the "optimal value penalty function" as

$$W^*(\epsilon, r) \equiv W(x(\epsilon, r), \epsilon, r) . \quad (2.11)$$

Armacost and Fiacco [2, Theorem 4 and Corollary 4.1] have obtained further results useful for estimating the first- and second-order sensitivity of the optimal value function  $f^*(\epsilon)$  of Problem  $P(\epsilon)$ .

Lemma 2.4 (Armacost and Fiacco [3]): If the assumptions A1 through A4 hold for Problem  $P(\epsilon)$ , then in a neighborhood of  $\epsilon=0$ ,

- (a)  $\lim_{r \rightarrow 0^+} W^*(\epsilon, r) = f^*(\epsilon)$  ;
- (b)  $\nabla_{\epsilon} W^*(\epsilon, r) = \nabla_{\epsilon} L(y, \epsilon)$  at  $y = y(\epsilon, r)$  ;
- (c)  $\lim_{r \rightarrow 0^+} \nabla_{\epsilon} W^*(\epsilon, r) = \nabla_{\epsilon} f^*(\epsilon)$  ;
- (d)  $\nabla_{\epsilon}^2 W^*(\epsilon, r) = \nabla_{\epsilon}^2 L(y(\epsilon, r), \epsilon)$  ; and
- (e)  $\lim_{r \rightarrow 0^+} \nabla_{\epsilon}^2 W^*(\epsilon, r) = \nabla_{\epsilon}^2 f^*(\epsilon)$  ;

where convergence is component by component in all cases.

Lemmas 2.1 through 2.4 enable us to calculate an estimate of  $y(\epsilon)$ ,  $\nabla_{\epsilon} y(\epsilon)$ ,  $\nabla_{\epsilon} f^*(\epsilon)$ , and  $\nabla_{\epsilon}^2 f^*(\epsilon)$  when  $\epsilon$  is near 0 and  $r$  is near 0, once  $y(\epsilon, r)$  is available.

In the next section we briefly present the algorithmic implementation of some of the above results.

## 2.2 The Algorithm Implementing the Sensitivity Results

The penalty function algorithm SUMT estimates the solution of the general mathematical problem  $P(\epsilon)$  by estimating the unconstrained minima

of the penalty function  $W(x, \epsilon, r)$  at successively decreasing values of the penalty function parameter  $r > 0$ . Under conditions weaker than those assumed here, Fiacco and McCormick [10] have shown that as  $r$  approaches zero, the sequence of the unconstrained minima of  $W(x, \epsilon, r)$  will approach a solution of  $P(\epsilon)$ . Each unconstrained penalty function minimization is thus a "subproblem" associated with a particular value of the penalty function parameter  $r$ .

The successive steps of the algorithm for first order sensitivity analysis of the Kuhn-Tucker triple with respect to the  $j$ th parameter that are implemented in SENSUMT [1] are listed below. Here the notation  $\bar{x}$  or  $\bar{x}(\bar{\epsilon})$  denotes the estimate of a solution point of  $P(\bar{\epsilon})$  calculated by SUMT for a given value of the penalty function parameter  $r$ , where  $\bar{\epsilon}$  denotes the value of the problem parameter at which this sensitivity is estimated.

Algorithm 2.1:

*Step 1.* Compute a representation of  $\nabla_x^2 W^{-1} = \nabla_x^2 W(\bar{x}, \bar{\epsilon}, r)^{-1}$  by L-U decomposition using the SUMT subroutines. If  $\nabla_x^2 W$  is not positive definite, terminate the sensitivity analysis.

*Step 2.* Estimate  $\partial(\nabla_x W^T)/\partial \epsilon_j$  using the central differencing formula

$$\partial(\nabla_x W^T)/\partial \epsilon_j \approx \left(\frac{1}{2\Delta}\right) \left( \nabla_x W(\bar{x}, \bar{\epsilon} + \Delta e_j, r)^T - \nabla_x W(\bar{x}, \bar{\epsilon} - \Delta e_j, r)^T \right),$$

where  $\Delta$  is the differencing interval and  $e_j$  is the  $j$ th unit vector.

*Step 3.* Calculate

$$\partial \bar{x}(\bar{\epsilon})/\partial \epsilon_j = -\nabla_x^2 W^{-1} \partial(\nabla_x W^T)/\partial \epsilon_j.$$

*Step 4.* Estimate  $\nabla_{\epsilon} g_1(\bar{x}, \bar{\epsilon})$  and  $\nabla_{\epsilon} h_1(\bar{x}, \bar{\epsilon})$  using

$$\partial g_1(\bar{x}, \bar{\epsilon})/\partial \epsilon_j \approx \left(\frac{1}{2\Delta}\right) \left( g_1(\bar{x}, \bar{\epsilon} + \Delta e_j) - g_1(\bar{x}, \bar{\epsilon} - \Delta e_j) \right), \text{ and}$$

$$\partial h_1(\bar{x}, \bar{\epsilon}) / \partial \epsilon_j \approx \left( \frac{1}{2\Delta} \right) (h_1(\bar{x}, \bar{\epsilon} + \Delta e_j) - h_1(\bar{x}, \bar{\epsilon} - \Delta e_j)) .$$

Step 5. Estimate the components of  $\nabla_{\epsilon} u(\bar{\epsilon})$  for  $i=1, \dots, m$  using

$$\partial u_i(\bar{\epsilon}) / \partial \epsilon_j \approx - \left( r / g_1(\bar{x}, \bar{\epsilon})^2 \right) \left( \nabla_x g_1(\bar{x}, \bar{\epsilon}) \partial \bar{x}(\bar{\epsilon}) / \partial \epsilon_j + \partial g_1(\bar{x}, \bar{\epsilon}) / \partial \epsilon_j \right) .$$

Step 6. Estimate the components of  $\nabla_{\epsilon} w(\bar{\epsilon})$  for  $i=1, \dots, p$  using

$$\partial w_i(\bar{\epsilon}) / \partial \epsilon_j \approx (1/r) \left( \nabla_x h_1(\bar{x}, \bar{\epsilon}) \partial \bar{x}(\bar{\epsilon}) / \partial \epsilon_j + \partial h_1(\bar{x}, \bar{\epsilon}) / \partial \epsilon_j \right) .$$

There are two methods for estimating  $\nabla_{\epsilon} f^*(\bar{\epsilon})$ , the first using  $\nabla_{\epsilon} f^* = \nabla_x f \nabla_{\epsilon} x + \nabla_{\epsilon} f$ , with  $\nabla_{\epsilon} x$  obtained from Step 3, and the second method using the gradient of the penalty function  $W$ , or equivalently, the Lagrangian taken with respect to the parameters [Lemma 2.1, conclusion (d); Lemma 2.4, conclusion (b)]. Both are incorporated in the computer program but used for different purposes. The former method gives the most accurate estimate of  $\nabla_{\epsilon} f^*(\bar{\epsilon})$  and is summarized in Steps 7 and 8.

Step 7. Estimate the components of  $\nabla_{\epsilon} f(\bar{x}, \bar{\epsilon})$  using the central differencing formula

$$\partial f(\bar{x}, \bar{\epsilon}) / \partial \epsilon_j \approx \left( \frac{1}{2\Delta} \right) (f(\bar{x}, \bar{\epsilon} + \Delta e_j) - f(\bar{x}, \bar{\epsilon} - \Delta e_j)) .$$

Calculate an estimate of the components of  $\nabla_{\epsilon} f^*(\bar{\epsilon})$  using the results of Steps 3 and 7 as

$$\partial f^*(\bar{\epsilon}) / \partial \epsilon_j \approx \nabla_x f(\bar{x}, \bar{\epsilon}) \partial \bar{x}(\bar{\epsilon}) / \partial \epsilon_j + \partial f(\bar{x}, \bar{\epsilon}) / \partial \epsilon_j .$$

The second method, using the gradient of the Lagrangian to estimate  $\nabla_{\epsilon} f^*(\bar{\epsilon})$ , is computationally less expensive and is used to obtain rough estimates that single out the more crucial parameters for further analysis. This approximation is calculated as follows.

Step 9. Estimate the components of  $V_{\epsilon} f^*(\bar{\epsilon})$  using the results of Steps 4 and 7 as

$$\begin{aligned} \partial f^*(\bar{\epsilon}) / \partial \epsilon_j &\approx \partial f(\bar{x}, \bar{\epsilon}) / \partial \epsilon_j - \sum_{i=1}^m u_i(\bar{\epsilon}, r) \partial g_i(\bar{x}, \bar{\epsilon}) / \partial \epsilon_j \\ &+ \sum_{i=1}^p w_i(\bar{\epsilon}, r) \partial h_i(\bar{x}, \bar{\epsilon}) / \partial \epsilon_j . \end{aligned}$$

### 3. Parametric Optimal Value Bounds

In this section we briefly review the procedure for calculating piecewise linear parametric upper and lower bounds on the optimal value function of convex problems with right hand side perturbations of the form  $CR(\epsilon)$ . It is important to note that the given technique is not only applicable to problems of this form, but also to any class of problems which have convex or concave optimal value functions. In addition, we show how the above technique can be extended to calculate similar bounds on the optimal value function of separable nonconvex right hand side perturbation problems  $SR(\epsilon)$ , once their (computable) over- and underestimating problems are available. For a more detailed treatment, readers should refer to [11].

#### 3.1 Parametric Bounds on the Optimal Value Function of Convex Right Hand Side Problems $CR(\epsilon)$

Consider the following right hand side parametric programming problem  $R(\epsilon)$ , discussed in Section 2:

$$\begin{aligned} &\text{minimize } f(x) \\ &\text{subject to } g_i(x) \geq \epsilon_i, \quad i=1, m \\ &\quad \quad \quad h_j(x) = \epsilon_j, \quad j=m+1, p \end{aligned} \quad R(\epsilon)$$

where  $x \in E^n$ ,  $f$ ,  $g$ , and  $h : E^n \rightarrow E^1$ , and  $C^2$  and  $\epsilon \in E^p$ . If  $f(x)$  and  $-g_i(x)$ ,  $i=1, m$ , are convex and  $h_j(x)$ ,  $j=m+1, p$  are linear, then

the problem  $R(\epsilon)$  is convex and will be designated by  $CR(\epsilon)$ . It is well known that  $f^*(\epsilon)$ , the optimal value function of the problem  $CR(\epsilon)$ , is a convex function of  $\epsilon$ .

The convexity of the optimal value function  $f^*(\epsilon)$  of the problem  $CR(\epsilon)$  enables one to calculate parametric upper and lower bounds on this function when any of the problem parameters is radically perturbed. This of course requires the solution and corresponding optimal value function sensitivity information for both perturbed and unperturbed problems. Before delving into the calculation procedure of these bounds, we must remind ourselves of the following two basic properties of convex functions.

- (i) Any line connecting two points on the graph of a convex function does not underestimate that function between the points.
- (ii) Any line tangent to the graph of a convex function does not overestimate that function.

These two properties lend themselves in a natural way to the calculation of parametric bounds on the optimal value function of the problem  $CR(\epsilon)$  under large perturbations of any of the problem parameters, say  $\epsilon_i$ .

### 3.2 Algorithm for Calculation of Bounds on $f^*(\epsilon)$ of the Problem $CR(\epsilon)$

In the following we list the successive steps required in calculation of bounds on  $f^*(\epsilon)$ , the optimal value function of  $CR(\epsilon)$ , as a function of  $\epsilon_i$ , when  $\epsilon_i$  is perturbed from  $\bar{\epsilon}_{i1}$  to  $\bar{\epsilon}_{i2}$  while the remaining parameters are fixed at their base values.

#### Algorithm 3.1:

*Step 1.* Solve the unperturbed problem and obtain  $f^*(\epsilon_i)$  and  $df^*(\epsilon_i)/d\epsilon_i$  at  $\epsilon_i = \bar{\epsilon}_{i1}$ . Note that



$$\frac{df^*(\epsilon_i)}{d\epsilon_i} = \begin{cases} u_i(\epsilon_i), & i=1, \dots, m \\ -w_i(\epsilon_i), & i=m+1, p. \end{cases} \quad (2.4)$$

$$(2.5)$$

Step 2. Re-solve the problem and obtain  $f^*(\epsilon_i)$  and  $df^*(\epsilon_i)/d\epsilon_i$  at  $\epsilon_i = \bar{\epsilon}_{i2}$ .

Step 3. Derive the equation of the line passing through the points  $(\epsilon_i = \bar{\epsilon}_{i1}, f^*(\epsilon_i) = f^*(\bar{\epsilon}_{i1}))$  and  $(\epsilon_i = \bar{\epsilon}_{i2}, f^*(\epsilon_i) = f^*(\bar{\epsilon}_{i2}))$ . This line provides a parametric upper bound  $\bar{f}^*(\epsilon_i)$  for  $f^*(\epsilon_i)$  as a function of  $\epsilon_i \in [\bar{\epsilon}_{i1}, \bar{\epsilon}_{i2}]$ .

Step 4. Derive the equation of the tangent lines to  $f^*(\epsilon_i)$  at the above two points with the slopes

$$\left. \frac{df^*(\epsilon_i)}{d\epsilon_i} \right|_{\epsilon_i = \bar{\epsilon}_{i1}} \quad \text{and} \quad \left. \frac{df^*(\epsilon_i)}{d\epsilon_i} \right|_{\epsilon_i = \bar{\epsilon}_{i2}}$$

calculated in Steps 1 and 2, respectively. These two lines provide a parametric lower bound  $\underline{f}^*(\epsilon_i)$  for  $f^*(\epsilon_i)$  as a function of  $\epsilon_i \in [\bar{\epsilon}_{i1}, \bar{\epsilon}_{i2}]$ .

The lines obtained in Steps 3 and 4 form a triangle which encloses the optimal value function  $f^*(\epsilon_i)$  over the given range of  $\epsilon_i$ . An estimate of  $f^*(\epsilon_i)$  can also be made by fitting a convex function that passes through the points given in Step 3 and having the corresponding slopes at these points obtained in Steps 1 and 2.

It is clear that the fundamental requirement in using the above algorithm is that the optimal value function be convex and differentiable at least at  $\epsilon_i = \bar{\epsilon}_{i1}$  and  $\epsilon_i = \bar{\epsilon}_{i2}$ . As it was given in Section 2, Fiacco, under the assumptions A1 - A4 of Lemma 2.1, has established the differentiability of the optimal value function in a neighborhood of  $\epsilon=0$  for the general parametric programming problem  $P(\epsilon)$ . Thus, not

only Problem CR( $\epsilon$ ), but all classes of parametric nonlinear programming problems possessing a convex optimal value function and, in particular, meeting assumptions A1 - A4 at  $\epsilon_i = \bar{\epsilon}_{i1}$  and  $\epsilon_i = \bar{\epsilon}_{i2}$ , can be analyzed for bounds according to Algorithm 3.1.

### 3.3 Parametric Bounds on the Optimal Value Function of the Problem SR( $\epsilon$ )

In this section we show how Algorithm 3.1 can be extended to calculate piecewise linear parametric upper and lower bounds on the optimal value function of a separable nonconvex right hand side perturbation problem, given the over- and underestimating problems.

#### 3.3.1 Separable right hand side perturbation problems.

The separable nonconvex problem addressed here has the following structure:

$$\begin{aligned} \text{minimize}_{x \in E^n} \quad & f(x) = \sum_{j=1}^n f_j(x_j) \\ \text{subject to} \quad & g_i(x) = \sum_{j=1}^n g_{ij}(x_j) \geq \epsilon_i, \quad i=1, \dots, m \quad \text{SR}(\epsilon) \\ & L_j \leq x_j \leq U_j, \quad j=1, \dots, n, \end{aligned}$$

where each  $f_j(x_j)$  and  $g_{ij}(x_j)$  is a function of a single variable  $x_j$  and is differentiable;  $L_j$  and  $U_j$  are lower and upper bounds on the variable  $x_j$ , respectively. Note that the discussion that follows is also valid when Problem SR( $\epsilon$ ) involves linear equality constraints. Let

$$\begin{aligned} G &\equiv \left\{ x : g_i(x) = \sum_{j=1}^n g_{ij}(x_j) \geq \epsilon_i, \quad i=1, \dots, m \right\}, \\ B_j &\equiv \{ x_j : L_j \leq x_j \leq U_j \}, \quad j=1, \dots, n, \\ B &\equiv \bigcap_{j=1}^n B_j, \end{aligned}$$

and

$$S \equiv G \cap B.$$

The set  $G$ , and therefore set  $S$ , depends on  $\epsilon$ . Thus the problem  $SR(\epsilon)$  will read as follows:

$$\begin{aligned} & \underset{x}{\text{minimize}} \quad f(x) = \sum_{j=1}^n f_j(x_j) \\ & \text{subject to} \quad x \in B \cap G \equiv S \subset E^n. \end{aligned} \quad SR(\epsilon)$$

### 3.3.2 Convex underestimating problem.

Let  $\tilde{f}_j(x_j)$  be the convex envelope of  $f_j(x_j)$ ,  $j=1, \dots, m$  and  $\tilde{g}_{ij}(x_j)$  be the concave envelope of  $g_{ij}(x_j)$ ,  $j=1, \dots, n$ ,  $i=1, \dots, m$  over the interval  $B_j$ . Thus, for  $x_j \in B_j$ ,

$$\tilde{f}_j(x_j) \leq f_j(x_j)$$

and

$$\tilde{g}_{ij}(x_j) \geq g_{ij}(x_j), \quad i=1, \dots, m; j=1, \dots, n.$$

Definition: Define the problem  $CSR(\epsilon)$  as follows:

$$\begin{aligned} & \underset{x \in E^n}{\text{minimize}} \quad \tilde{f}(x) = \sum_{j=1}^n \tilde{f}_j(x_j) \\ & \text{subject to} \quad x \in G \cap B \equiv \tilde{S}, \end{aligned}$$

where

$$\tilde{G} \equiv \left\{ x : \tilde{g}_i(x) = \sum_{j=1}^n \tilde{g}_{ij}(x) \geq \epsilon_i, \quad i=1, \dots, m \right\}.$$

It follows easily that problem  $CSR(\epsilon)$  is a separable convex RHS perturbation problem whose optimal value function underestimates the global optimal value function of the problem  $SR(\epsilon)$ .

Thus the basic ingredients required to formulate the convex underestimating problem  $CSR(\epsilon)$  are the convex envelope of each term in the

objective function and the concave envelope of each term in the constraints of the problem  $SR(\epsilon)$  over the rectangular polytope  $B$ .

In the next section we will discuss the formulation of a convex overestimating problem  $C\tilde{S}R(\epsilon)$  of the problem  $SR(\epsilon)$ .

### 3.3.3 Convex overestimating problem.

Let  $\tilde{f}_j(x_j)$  be a convex function which does not underestimate  $f_j(x_j)$ ,  $j=1, \dots, n$ , and  $\tilde{g}_{ij}(x_j)$  be a concave function which does not overestimate  $g_{ij}(x_j)$ ,  $j=1, \dots, n$ ,  $i=1, \dots, m$ , over the interval  $B_j$ . Thus, for  $x_j \in B_j$ ,

$$\tilde{f}_j(x_j) \geq f_j(x_j)$$

and

$$\tilde{g}_{ij}(x_j) \leq g_{ij}(x_j), \quad i=1, \dots, m; j=1, \dots, n.$$

Henceforth in discussion,  $\tilde{f}_j(x_j)$  will be abbreviated as convex "nuf" (nonunderestimating function) of the single variable function  $f_j(x_j)$ , and  $\tilde{g}_{ij}(x_j)$  will be abbreviated as concave "nof" (nonoverestimating function) of  $g_{ij}(x_j)$  over the interval  $B_j$ ,  $i=1, \dots, m$ ,  $j=1, \dots, n$ .

Definition: Let us define the problem  $C\tilde{S}R(\epsilon)$  as follows:

$$\begin{aligned} &\underset{x \in E^n}{\text{minimize}} \quad \tilde{f}(x) = \sum_{j=1}^n \tilde{f}_j(x_j) \\ &\text{subject to} \quad x \in (\tilde{G} \cap B) \equiv \tilde{S}, \end{aligned}$$

where

$$\tilde{G} \equiv \left\{ x : \tilde{g}_i(x) = \sum_{j=1}^n \tilde{g}_{ij}(x_j) \geq \epsilon_i, \quad i=1, \dots, m \right\}.$$

It is easy to show that problem  $C\tilde{S}R(\epsilon)$  is a separable convex RHS problem whose optimal value function underestimates the global optimal value function of the problem  $SR(\epsilon)$ .

Thus, in order to formulate an overestimating convex problem  $\tilde{CSR}(\epsilon)$  of the problem  $SR(\epsilon)$ , one must calculate convex nuf,  $f_j(x_j)$ , for each component of  $f(x)$  and concave nuf,  $\tilde{g}_{ij}(x_j)$ , for each component of  $g_i(x)$ ,  $i=1, \dots, m$ , over the interval  $B_j$ . It must be noted that convex nuf and concave nuf of a given function, unlike its envelopes, are not unique. Thus the overestimating problem  $\tilde{CSR}(\epsilon)$  of problem  $SR(\epsilon)$  is not unique. For further treatment of convex nuf and concave nuf of single variable functions, readers may refer to [11].

An important point to be noted here is that formulation of the convex problems  $\tilde{CSR}(\epsilon)$  and  $CSR(\epsilon)$  enable one to calculate parametric upper and lower bounds on the (global) optimal value function of the separable nonconvex programming problem  $SR(\epsilon)$ .

### 3.4 Algorithm for Calculation of Bounds on the (Global) Optimal Value Function of the Problem $SR(\epsilon)$

In the following an algorithm is presented for calculation of upper and lower bounds on the (global) optimal value function of the problem  $SR(\epsilon)$  as a function of each RHS parameter  $\epsilon_i$ , where  $\epsilon_i \in [\bar{\epsilon}_{i1}, \bar{\epsilon}_{i2}]$ . Implicit in this algorithm is the assumption that the overestimating problem  $\tilde{CSR}(\epsilon)$  has a nonempty interior, which is a basic requirement of almost any nonlinear programming algorithm.

#### Algorithm 3.2:

*Step 1:* Calculate  $f_j(x_j)$ , a convex nuf for each  $f_j(x_j)$ ,  $j=1, \dots, n$ , and  $\tilde{g}_{ij}(x_j)$ , a concave nuf for each  $g_{ij}(x_j)$ ,  $i=1, \dots, m$ ,  $j=1, \dots, n$ , in closed interval  $B_j$ , and construct the corresponding overestimating problem  $\tilde{CSR}(\epsilon)$ .

*Step 2:* Solve the problem  $\tilde{CSR}(\epsilon)$  at  $\epsilon_i = \bar{\epsilon}_{i1}$  and  $\epsilon_i = \bar{\epsilon}_{i2}$  and obtain  $f^*(\epsilon_i)$  at these values of  $\epsilon_i$ .

*Step 3:* Derive the equation for the line passing through the points  $(\epsilon_i = \bar{\epsilon}_{i1}, f^*(\epsilon_i) = f^*(\bar{\epsilon}_{i1}))$  and  $(\epsilon_i = \bar{\epsilon}_{i2}, f^*(\epsilon_i) = f^*(\bar{\epsilon}_{i2}))$ . This line provides a parametric upper bound  $\bar{f}^*(\epsilon_i)$  on the (global) optimal value function  $f^*(\epsilon_i)$  of the problem  $SR(\epsilon)$  as a function of  $\epsilon_i$  over the interval  $[\bar{\epsilon}_{i1}, \bar{\epsilon}_{i2}]$ .

*Step 4:* Calculate  $\bar{f}_j(x_j)$ , the convex envelope for each  $f_j(x_j)$ ,  $i=1, \dots, n$  and  $\bar{g}_{ij}(x_j)$ , the concave envelope for each  $g_{ij}(x_j)$ ,  $i=1, \dots, m$ ,  $j=1, \dots, n$  in closed interval  $B_j$ , and construct the corresponding underestimating problem  $CSR(\epsilon)$ .

*Step 5:* Solve the problem  $CSR(\epsilon)$  at  $\epsilon_i = \bar{\epsilon}_{i1}$  and  $\epsilon_i = \bar{\epsilon}_{i2}$  and obtain  $\bar{f}^*(\epsilon_i)$  and  $\bar{u}_i(\epsilon_i) = d\bar{f}^*(\epsilon_i)/d\epsilon_i$  at these values of  $\epsilon_i$ .

*Step 6:* Derive the equation for the tangent lines to  $\bar{f}^*(\epsilon_i)$  at  $\epsilon_i = \bar{\epsilon}_{i1}$  and  $\epsilon_i = \bar{\epsilon}_{i2}$ , with respective slopes of  $\bar{u}_i(\bar{\epsilon}_{i1})$  and  $\bar{u}_i(\bar{\epsilon}_{i2})$  derived in Step 5. These two lines over the interval  $[\bar{\epsilon}_{i1}, \bar{\epsilon}_{i2}]$  jointly provide a parametric lower bound  $\underline{f}^*(\epsilon_i)$  on the (global) optimal value function  $f^*(\epsilon_i)$  of the problem  $SR(\epsilon)$  as a function of  $\epsilon_i$ .

The implicit assumption in Steps 5 and 6 of this algorithm is that assumptions A1 - A4 of Lemma 2.1 hold at the solution points when  $\epsilon_i$  takes values of  $\bar{\epsilon}_{i1}$  and  $\bar{\epsilon}_{i2}$ .

Before delving into the next section, we must point out that if over- and underestimating problems with concave optimal value functions could be formulated, then the algorithm above, with minor alterations, can be used to calculate the desired bounds. In this instance Step 6, when applied to the overestimating problem, will provide an upper bound

while Step 3, when applied to the underestimating problem, will provide a lower bound on the optimal value function. For calculation of envelopes, convex nuf and concave nof of single variable functions, readers may refer to [11]. The entries in Table 1 summarize the concept of convex nuf and concave nof for a variety of single variable functions.

#### 4. SENSUMT Input Specifications

The coding procedure for nonlinear programming problems for solution, sensitivity analysis, and bound calculation using SENSUMT closely follows the coding procedure for SUMT-Version 4 [13]. In order to code a given problem for the above calculations, one must have (i) a user-supplied subroutine, and (ii) user's information cards.

##### 4.1 User-supplied Subroutines

User-supplied subroutines will generally consist of four subroutines called READIN, RESTNT, GRAD1 and MATRIX. These subroutines are the only subroutines for SENSUMT that are problem dependent. The communication between these subroutines and the rest of the subroutines of SENSUMT is mostly through the COMMON blocks, but partly via their arguments. Each of the user-supplied subroutines must contain the double precision card

IMPLICIT REAL\*8(A-H,O-Z).

The subroutines RESTNT, GRAD1, and MATRIX must contain the COMMON card

COMMON/SHARE/X(45), DEL(45), A(45,45), N, M, MN, NP1, NM1.

Also, depending on the problem being solved, the following COMMON card may be required in some or all of the user subroutines:

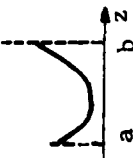

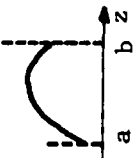

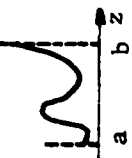
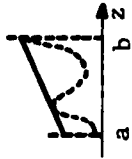
COMMON/SEN/PAR(45), DPAR(45), NPAR, ISENS.

When bound analysis is desired, subroutine READIN must contain the COMMON card

COMMON/ABG/LY, LZ, PER(45).

TABLE 1

CONVEX nuf AND CONCAVE nuf OF THE SINGLE VARIABLE FUNCTION  $T(z)$  OVER THE CLOSED INTERVAL  $a \leq z \leq b$

$T(z)$	$T(z)$ [convex nuf of $T(z)$ ] $a \leq z \leq b$	$\tilde{T}(z)$ [concave nuf of $T(z)$ ] $a \leq z \leq b$
Graph	Graph	Graph
Convex	 $T(z) = \tilde{T}(z)$	 $\tilde{T}(z) = \text{any line not overestimating } T(z)$
Concave	 $T(z) = \tilde{T}(z)$	 $\tilde{T}(z) = \text{any line not underestimating } T(z)$
Neither convex nor concave	 $T(z) = \tilde{T}(z)$	 $\tilde{T}(z) = \text{any line not overestimating } T(z)$



In the following section we discuss the purpose and coding procedure of each of these user-supplied subroutines. The general description and listing of the rest of the subroutines comprising SENSUMT will be given in Section 6.

#### 4.1.1 READIN

This subroutine allows the user to read in the data which is necessary for evaluation of the problem functions and convey it, through the COMMON blocks, to the related subroutines. The problem parameters for which sensitivity and bound calculation are desired must also be defined and initialized in this subroutine. The initial values of the parameters must be placed in array PAR. The corresponding perturbation values for bound calculation must be placed in array PER. NPAR in this subroutine must be initialized to the total number of problem parameters on which sensitivity is desired. *Notice that the presence of array PER in this subroutine will trigger SENSUMT to calculate the desired bound calculation.* READIN may also be used to read in and print out any pertinent information of the user's choice, such as problem title, date, and so on. This subroutine is called only once for each problem being solved. For clarity, see Example 1 and its code in Figure 2. Any input data read by subroutine READIN must be placed immediately after the first option card.

#### 4.1.2 RESTNT (IN,VAL)

This subroutine is used to read in the problem functions. For  $IN=0$ , set  $VAL=f(x)$  and for  $IN \neq 0$  set

$$VAL = g_{IN}(x) , \quad IN=1,2,\dots,m$$

$$VAL = h_{IN}(x) , \quad IN=m+1,\dots,m+p .$$

For clarity see Example 1 and Figure 2.

## 4.1.3 GRAD1 (IN)

This subroutine is used to read in the gradients of the problem functions. For IN=0 set

$$\text{DEL}(J) = f(x)/\partial x_J, \quad J=1,2,\dots,n$$

and for IN≠0 set

$$\text{DEL}(J) = \begin{cases} \partial g_{\text{IN}}(x)/\partial x_J, & J=1,2,\dots,n \\ \partial h_{\text{IN}}(x)/\partial x_J, & J=1,2,\dots,n \end{cases}$$

For an illustration, see Example 1 and Figure 2.

SENSUMT can internally compute numerical approximations for some or all of the gradients. To utilize this option, instead of coding the gradients code the statements

CALL DIFF1(IN)

RETURN

for any desired value of the variable IN (the index of the problem functions, with IN=0 corresponding to the objective function  $f(x)$ ). Because of the computational effort involved in numerical differencing, this option may be prohibitive for large problems.

## 4.1.4 MATRIX (IN,K)

This subroutine reads in the *upper triangle and diagonal elements* of the Hessian matrix of the problem functions. For IN=0, set

$$A(I,J) = \partial^2 f(x)/\partial x_I \partial x_J, \quad I=1,2,\dots,n; J=1,2,\dots,n; \text{ and } I \leq J.$$

For IN≠0 set

$$A(I,J) = \begin{cases} \partial^2 g_{\text{IN}}(x)/\partial x_I \partial x_J, & \text{IN}=1,2,\dots,m \\ \partial^2 h_{\text{IN}}(x)/\partial x_I \partial x_J, & \text{IN}=m+1,\dots,p \end{cases}$$

where  $I=1,2,\dots,n$ ;  $J=1,2,\dots,n$ ; and  $I \leq J$ . Before the call is made to this subroutine, all entries of the matrix  $A(I,J)$  are set to zero.

The second argument  $K$  of the subroutine MATRIX is provided so that the user may communicate to SENSUMT that the Hessian of a problem function is identically zero. Set  $K=1$  if the second partial derivatives of the  $IN$ th constraint are zero ( $IN=0$  corresponds to the objective function). For an illustration, see Example 1 and Figure 2.

SENSUMT can internally compute numerical approximations for some or all of the Hessian matrices. To utilize this option, instead of coding the Hessian matrices, code the statements

```
CALL DIFF2(IN)
RETURN
```

for any desired value of the variable  $IN$  (the index of the problem function, with  $IN=0$  designating the objective function). *Although this option can be used in calculating an optimal solution, the current sensitivity routines require explicit coding of the closed form of the Hessian matrices. Thus, this option cannot be used for sensitivity and bound calculations.* Again, because of the computational effort involved in numerical differencing, use of this option is not advised for large problems.

#### 4.2 User's Information Cards

The other inputs required by SENSUMT are the user's information cards, i.e.,

- PARAMETER CARD
- INITIAL VECTOR CARD(S)
- FIRST OPTION CARD
- TOLERANCE CARD
- SECOND OPTION CARD.

In the following we specify the type of information, along with the corresponding format, name, and description, which must be provided by the user to SENSUMT via the above cards. All of this information is read by the program MAIN, which sets up the SENSUMT algorithm to solve and

analyze the problem under study. Tables 2, 3, and 4, which follow, are taken from [13].

#### 4.2.1 Parameter card.

Input specifications of the parameter card are shown in Table 2.

TABLE 2  
PARAMETER CARD

Columns	Format	Name	Use
01-12	E12.0	EPSI ( $\epsilon$ )	Tolerance used to decide if an unconstrained minimum has been achieved for each subproblem [see Option 9]
13-24	E12.0	RH0IN ( $r_1$ )	Possible initial value of $r$ (often set at 1.0) [see Option 1]
25-36	E12.0	THETA0 ( $\theta_0$ )	Tolerance used to decide if the solution to the NLP problem $P(\epsilon)$ has been approximated [see Option 5]
37-48	E12.0	RATIO ( $c$ )	Parameter ( $>1$ ) used to compute consecutive values of $r$ ; $r_{i+1} = r_i/c$ (often set at 16.0)
49-60	E12.0	TMMAX	Maximum amount of time for solving problem (in seconds)
61-64	I4	M	Number (integer) of nontrivial constraints [see Option 2] $(M+MZ) \leq 200^*$
65-68	I4	N	Number (integer) of variables, $N \leq 45$
69-72	I4	MZ	Number (integer) of equality constraints

\*The limits on  $M+MZ$  and  $N$  are governed by the size of the arrays in SENSUMT.

#### 4.2.2 Initial vector card(s).

The cards designating the initial starting point immediately follow the parameter card. There are six components per card, requiring  $n/6$  cards for the initial vector. Each card has the format 6E12.6.

#### 4.2.3 First option card.

The input specifications for the first option card are shown in Table 3. This card must immediately follow the user-supplied subroutines.

#### 4.2.4 Tolerance card.

The input specifications for the tolerance card are given in Table 4. This card must immediately follow the first option card.

#### 4.2.5 Second option card.

The input specifications for the second option card, which immediately follows the tolerance card, are given in Table 5.

Figure 1 shows the arrangement of the data together with JOB and JCL cards in a coded deck.

TABLE 3  
FIRST OPTION CARD

Option	Column	Value	Meaning
1 (normally set to 3)	7	=1	The value for $r$ is made by finding an approximate solution $\min\{[\nabla W(x^0, r)[\nabla^2 W(x^0, r)]^{-1}\nabla W(x^0, r)\}$ which is a good approximation only when $x^0$ is close to the boundary (i.e., for some $i$ , $g_i(x)$ is close to zero) or when $\nabla^2 f(x^0)=0$ and when $MZ=0$ .
		=2	$r_1$ is given by formula 8.65 [10, p. 191] (can only be used when $MZ=0$ ).
		=3	$r_1 = RH\emptyset IN$ (see parameter card).
2	14	=1	The requirements (trivial constraints) that $x_i \geq 0$ for $i=1, \dots, n$ are to be automatically included in the problem.
		=2	The only constraints on the problem are those inputted by the user.
3 (normally set to 1)	21	=1	Standard printout (this includes a call to OUTPUT after the solution of every subproblem). Also the estimates of the "Lagrange multipliers" and first and second order solution estimates are printed.
		=2	For additional printout (includes standard printout and every intermediate point, gradient of $P$ and the vector $S$ ).
4 (normally set to 1)	28	=1	Final convergence is determined on the basis of current solution to the subproblem.
		=2	Final convergence is determined on the basis of the first order estimates. The first order estimate of the solution vector must be close to feasible. See below.
		=3	Final convergence is determined on the basis of the second order estimates. The second order estimate of the solution vector must be close to feasible before the convergence check is made. If $\bar{x}$ is a solution

Option	Column	Value	Meaning
5 (normally set to 1) (see [10, p. 195])	35		estimate it is considered close to being feasible if $g_i(\bar{x}) + \theta_0 \geq 0$ , $i=1,2,\dots,m$ , where $\theta_0$ is defined on the parameter card.
		=1	The convergence criterion determining the NLP problem has been solved (only use =1, when NT4#1).
		=2	Quit when $\frac{G - f[x(r_k)]}{G[x(r_k), \mu(r_k), \lambda(r_k)]} < \theta_0$
		=3	Quit when $ r \sum_{j=1}^m \ln g_j[x(r_k)]  < \theta_0$
		=3	Quit when $\frac{\text{first order estimate of } v_0}{G[x(r_k), \mu(r_k), \lambda(r_k)]} - 1 < \theta_0$
6 (normally set to 1)	42	=1	After final convergence the program reads in new data and solves the next problem.
		=2	After final convergence has been determined a call to PUNCH is made before proceeding on to the next problem.
7 (normally set to 1)	49		First move after a minimum to a subproblem is achieved
		=1	No extrapolation.
		=2	Extrapolate through last two minima.
		=3	Extrapolate through last three minima.
8	56		Not used.
9	63		Subproblem convergence criterion, or when to stop minimizing P function for fixed value of r (see parameter card).
		=1	Quit when $ \nabla_x W^T(x^i, r) \left[ \frac{\partial^2 W(x, r)}{\partial x_i \partial x_j} \right]^{-1} \nabla_x W(x^i, r)  < \epsilon$ .
		=2	Quit when $ \nabla_x W^T(x^i, r) \left[ \frac{\partial^2 W(x, r)}{\partial x_j \partial x_j} \right]^{-1} \nabla_x W(x^i, r)  < \frac{W(x^{i-1}) - W(x^i)}{5}$ .

Table 3--continued

Option	Column	Value	Meaning
10 <sup>a</sup>	70	=3	Quit when $ \nabla_x W(x^i, r)  < \epsilon$
		=1	At least one nonlinear constraint
		=2	Linear constraints
		=3	Linear constraints and linear objective function (i.e., a linear programming problem)

<sup>a</sup>When option 10=3, MATRIX (the user subroutine supplying the second partial derivatives) will not be called, and when option 10=2 it will be called only to get the second partials of  $f(x)$ .

TABLE 4

## TOLERANCE CARD

Columns	Format	Name	Description
01-12	E12.6	XEP1	If some first or second derivatives are to be gotten by numerical differencing (see addition option card) this is the value used to compute them. See the description of DIFF1. (Usually setting XEP1 equal to .0001 is satisfactory, although a good value is dependent on the scaling of the problem.)
13-24	E12.6	XEP2	When minimizing the W function for a given value of $r$ (RH0) the value of W must decrease by an amount exceeding XEP2 for each iteration after the first. If it does not, then the code prints out the message "apparently roundoff errors prevent a more accurate determination of the minimum of this subproblem," and it is assumed a minimum has been found. (Usually we set XEP2 equal to 0.)



TABLE 5  
SECOND OPTION CARD

Option	Column	Value	Meaning
1 (normally set to 4)	7	=1	Solve problems without checking derivatives
		=2	Solve problem after checking only first derivatives
		=3	Do not solve problem after checking only first derivatives
		=4	Solve problem after checking first and second derivatives
		=5	Do not solve problem after checking first and second derivatives
2	14	=1	The method for minimizing the unconstrained penalty function is to be the generalized Newton-Raphson method as modified to handle indefinite Hessian matrices. This method requires function values, first and second derivatives.
		=2	Same as 1, except that when an "orthogonal move" is made because of an indefinite Hessian matrix, $-\nabla P$ is added to the orthogonal move vector.
		=3	Steepest descent is used to minimize P-function.
		=4	The method for minimizing the unconstrained penalty function is McCormick's modification of the Fletcher-Powell method as reported in [10]. This requires function values and first derivatives.
3	21	=0	Do not conduct a sensitivity analysis.
		=1	Conduct a sensitivity analysis at the final subproblem with a fixed value for the differencing interval.
		=2	Conduct a sensitivity analysis at each subproblem along the minimizing trajectory with the value of the differencing interval depending on the particular subproblem.
		=3	Conduct a sensitivity analysis at the final subproblem for a range of differencing intervals.

Table 5--continued

Option	Column	Value	Meaning
4	28	=0	Do not estimate the partial derivatives of the estimates of the Lagrange multipliers.
		=1	Estimate the partial derivatives of the estimates of the Lagrange multipliers whenever a sensitivity analysis of the solution point is conducted.
5	35	=0	Estimate the partial derivatives of the optimal value function and eliminate those parameters which do not affect the optimal value functions from subsequent sensitivity calculations.
		=1	Estimate the partial derivatives of the optimal value function with respect to all parameters, but continue all subsequent sensitivity calculations with respect to all parameters.
		=2	Do not estimate the partial derivatives of the optimal value function first. Conduct the sensitivity analysis with respect to all parameters.
6	42	=0	Do not transform the results.
		=1	The problem being solved, $P(x)$ , is a convex equivalent of a geometric programming problem $G(t)$ obtained by transformation $c_i = e^{-xi}$ . Thus back transform the results to $t$ space.
7	49	=1	$f^*(\epsilon)$ is convex. Derive parametric upper and lower bounds on $f^*(\epsilon)$ .
		=2	$f^*(\epsilon)$ is convex. Derive parametric upper bound on $f^*(\epsilon)$ .
		=3	$f^*(\epsilon)$ is convex. Derive parametric lower bound on $f^*(\epsilon)$ .
		=4	$f^*(\epsilon)$ is concave. Derive parametric upper and lower bounds on $f^*(\epsilon)$ .
		=5	$f^*(\epsilon)$ is concave. Derive parametric upper bound on $f^*(\epsilon)$ .
		=6	$f^*(\epsilon)$ is concave. Derive parametric lower bound on $f^*(\epsilon)$ .

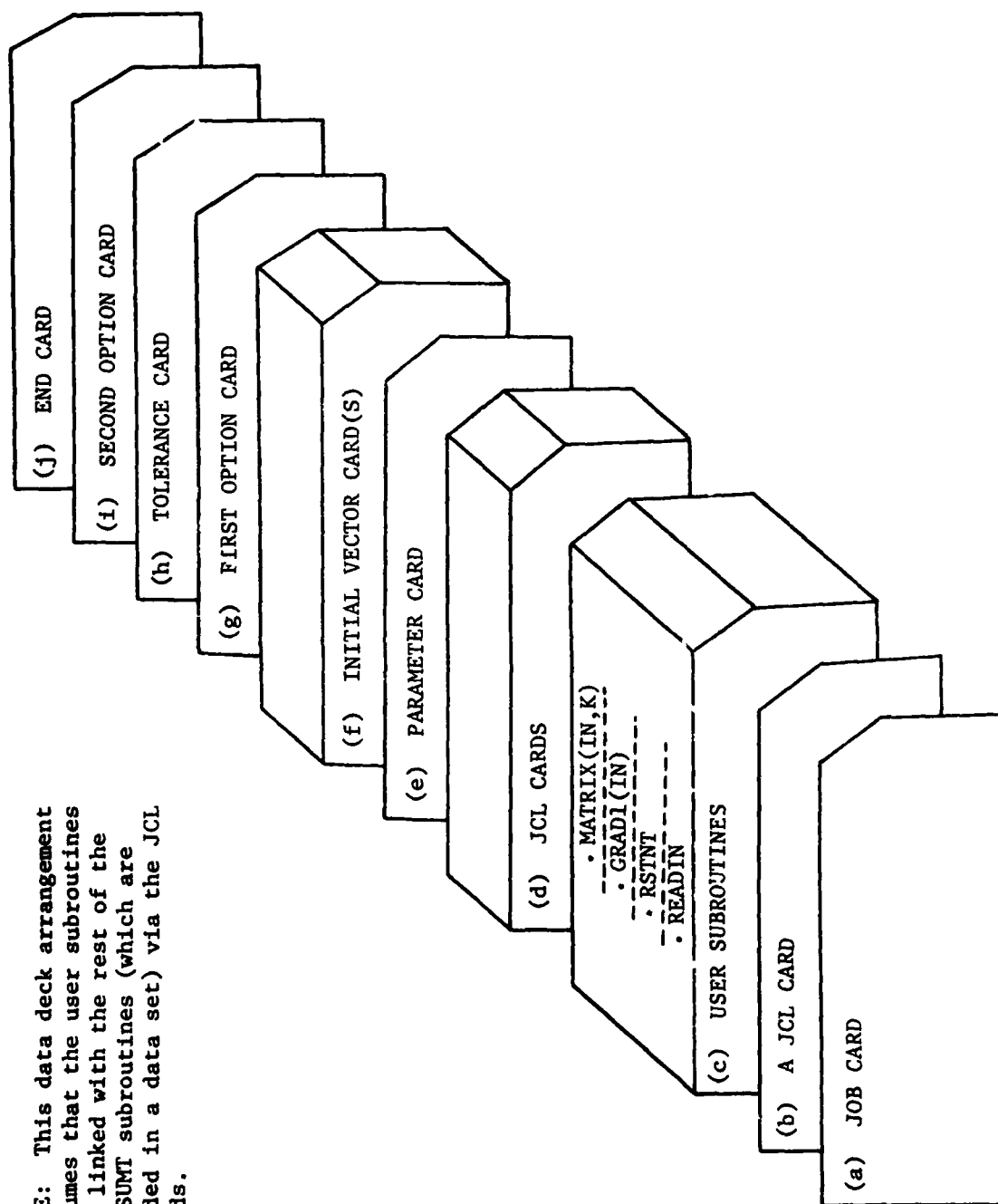


Figure 1.--Data deck structure for SENSUMT.

## 5. Coded Examples and Input/Output Illustrations

In this section we provide the input listing and output description of two illustrative examples which are taken from [11].

### 5.1 Example 1

Consider the following convex RHS programming problem:

$$\begin{aligned} \text{minimize } f(x) &= (x_1 - 4)^2 + (x_2 - 2)^2 \\ \text{subject to } g_1(x) &= -x_1^2 + x_2 \geq \epsilon_1 \\ g_2(x) &= -x_1 - x_2 \geq \epsilon_2 . \end{aligned}$$

It is desired to solve and analyze this problem for sensitivity when  $\epsilon_1 = 0$  and  $\epsilon_2 = -3$ . Moreover, it is desired to derive parametric upper and lower bounds on the optimal value function of this problem when

- (i)  $-1 \leq \epsilon_1 \leq 0$  while  $\epsilon_2 = -3$ , and
- (ii)  $-3 \leq \epsilon_2 \leq -1$  while  $\epsilon_1 = 0$ .

#### 5.1.1 Computer listing of the code deck.

Figure 2 shows the computer listing of the code of Example 1. The letters in circles categorizing the input data correspond to those indicated in the code deck structure depicted in Figure 1. The format of the listed data is given in Section 4.

#### 5.1.2 Selected pages from the computer output.

Figure 3 lists an annotated computer output for Example 1. An explanation of the meaning of the corresponding steps follows.

J=GHAEML T=3 I=3	① Job card
// EXEC FORG1	② A JCL card
SUBROUTINE READIN IMPLICIT REAL*8(A-H,O-Z) COMMON/SEN/PAR(45),DPAR(45),NPAR,ISENS COMMON/ABG/LY,LZ,PER(45) NPAR=2 PAR(1)=0. PAR(2)=3.0 PER(1)=1. PER(2)=2. RETURN END	③ User subroutines
SUBROUTINE RSTNT(IN,VAL) IMPLICIT REAL*8(A-H,O-Z) COMMON/SHARE/X(45),DEL(45), A(45,45),N,M,MN,NPI,NMI COMMON/SEN/PAR(45),DPAR(45),NPAR,ISENS COMMON/ABG/LY,LZ,PER(45) I=IN+1 GO TO (20,1,2),I 20 VAL=(X(1)-4.)*2+(X(2)-2.)*2 RETURN 1 VAL=-X(1)+2+X(2)-PAR(1) RETURN 2 VAL=-X(1)-X(2)-PAR(2) RETURN END	Subroutine READIN Subroutine RSTNT (IN,VAL)
SUBROUTINE GRADT(IN) IMPLICIT REAL*8(A-H,O-Z) COMMON/SHARE/X(45),DEL(45), A(45,45),N,M,MN,NPI,NMI COMMON/SEN/PAR(45),DPAR(45),NPAR,ISENS J=IN+1 DO 5 I=1,N DEL(I)=0.0 5 CONTINUE GO TO (20,1,2),J 20 DEL(1)=2.*(X(1)-4.) DEL(2)=2.*(X(2)-2.) RETURN 1 DEL(1)=-2.*X(1) DEL(2)=1.0 RETURN 2 DEL(1)=-1.0 DEL(2)=-1.0 RETURN END	Subroutine GRAD (IN)
SUBROUTINE MATRIX(IN,K) IMPLICIT REAL*8(A-H,O-Z) COMMON/SHARE/X(45),DEL(45), A(45,45),N,M,MN,NPI,NMI COMMON/SEN/PAR(45),DPAR(45),NPAR,ISENS L=IN+1 GO TO (20,1,2),L 20 A(1,1)=2.0 A(1,2)=0. A(2,2)=2.0 RETURN 1 A(1,1)=-2.0 A(1,2)=0. A(2,2)=0. RETURN 2 K=1 RETURN END	Subroutine MATRIX (IN,K)
// EXEC FORG6,DSN=OR799661.BOUNDS45',PROG=MAIN //GO,SYSLIB DD // DD // DD DSN=GNU.FORTLIB,DISP=SHR //GO,SYSLIN DD // DD DSN=&TEMPUNCH,DISP=(OLD,DELETE) //F107F001 DD SYS.WT=A	④ JCL cards
1.0E-08 10.0 1.0E-08 4.0 900.0 2 2 0	⑤ Parameter card
3 2 1 1 1 1 1 1	⑥ Initial vector card
0.1E-10 0.00 1 1 0 1	⑦ First option card
	⑧ Tolerance card
	⑨ Second option card
//	⑩ End card

Figure 2.--Computer listing of the code for Example 1.

```

1  NONLINEAR PROGRAMMING - JTIME-SUMT VERSION 4 03/01/73
2  N= 2  K= 2  MZ= 0
3  MAX. TIME= 3.900000E 03  R= 0.1000000E 02  RATIO= 0.4000000E 01  EPSILON= 0.9999997E-08  THETA= 0.9999996E-06
4  OPTIMIZATION SELECTED 1 1 1 1 1 0 1 1 1
5  TOLERANCES 0.0 0.9999999E-03
6  SECOND SET OF OPTIONS 1 1 1 0 1
7  TIME= 0.059 SECONDS P= 0.0 G= 0.0 RSIGMA= 0.0 H= 0.0
8  THE CURRENT VALUE OF X IS 0.1000000 01
9  THE CONSTRAINT VALUES 0.1500000 01
10  0.5000000 00 0.5000000 00
11  TIME= 0.102 SECONDS
12  *****THE FEASIBLE STARTING POINT TO BE USED IS ***
13  F= 0.9250000 01 P= 0.0 G= 0.0 RSIGMA= 0.0 H= 0.0
14  THE CURRENT VALUE OF X IS 0.1000000 01 0.1500000 01
15  THE CONSTRAINT VALUES 0.5000000 00 0.5000000 00
16  APPARENTLY ROUNDOFF ERRORS PREVENT A MORE ACCURATE DETERMINATION OF THE MINIMUM OF THIS SUBPROBLEM.
17  TIME= 0.219 SECONDS
18  *****
19  POINT= 4 QDIT= 0.4061930-06 RHO= 0.1000000 02 MAGNITUDE= 0.19361530-02 PHASE= 2
20  F= 0.15780290 02 P= 0.80717330 01 G= -0.42197120 01 RSIGMA= -0.77085550 01 H= 0.0
21  THE CURRENT VALUE OF X IS 0.51007220-01 0.15690190 01
22  THE CONSTRAINT VALUES 0.6283900 01 0.72465150 01
23  APPARENTLY ROUNDOFF ERRORS PREVENT A MORE ACCURATE DETERMINATION OF THE MINIMUM OF THIS SUBPROBLEM.
24  TIME= 0.340 SECONDS
25  *****
26  LAGRANGE MULTIPLIERS
27  F= 0.15780290 02 P= 0.80717330 01 G= -0.42197120 01 RSIGMA= -0.77085550 01 H= 0.0
28  THE CURRENT VALUE OF X IS 0.51007220-01 0.15690190 01
29  THE CONSTRAINT VALUES 0.6283900 01 0.72465150 01
30  APPARENTLY ROUNDOFF ERRORS PREVENT A MORE ACCURATE DETERMINATION OF THE MINIMUM OF THIS SUBPROBLEM.
31  TIME= 0.340 SECONDS
32  *****
33  POINT= 7 QDIT= 0.23087850-04 RHO= 0.2500000 01 MAGNITUDE= 0.18934470-01 PHASE= 2
34  F= 0.11111590 02 P= 0.11665690 02 G= 0.61115940 01 RSIGMA= 0.55609590 00 H= 0.0
35  THE CURRENT VALUE OF X IS 0.69826920 00 0.15615600 01
36  THE CONSTRAINT VALUES 0.10539800 01 0.76017110 00
37  1ST ORDER ESTIMATES
38  F= 0.97418970 01 P= 0.12863680 02 G= 0.05553620 01 RSIGMA= 0.0 H= 0.0
39  THE CURRENT VALUE OF X IS 0.91402710 02 0.15124070 01
40  THE CONSTRAINT VALUES 0.00000000 00 0.00000000 00

```

Figure 3.--Annotated computer output for Example 1.

LAGRANGE MULTIPLIERS  
 P= 0.9741897D 01 P= 0.1288368D 02 G= 0.9555362D 01 RSIGNA= 0.0 H= 0.0  
 THE CURRENT VALUE OF X IS  
 0.9140231D 00 0.1532407D 01  
 THE CONSTRAINT VALUES  
 0.2371962D 01 0.3288733D 01  
 APPARENTLY ROUND-OFF ERRORS PREVENT A MORE ACCURATE DETERMINATION OF THE MINIMUM OF THIS SUBPROBLEM.  
 TIME= 0.469 SECONDS

12 POINT= 10 DOUT= 0.7459573D-04 PHASE= 2  
 F= 0.8911950D 01 P= 0.9906426D 01 RHO= 0.4250000D 00 MAGNITUDE= 0.4996642D-01 H= 0.0  
 THE CURRENT VALUE OF X IS  
 0.1106835D 01 0.1623767D 01 RSIGNA= 0.1394676D 01 H= 0.0  
 THE CONSTRAINT VALUES  
 0.3986811D 00 0.2693975D 00

2ND ORDER ESTIMATES  
 F= 0.7596676D 01 P= 0.9083760D 01 G= 0.7645402D 01 RSIGNA= 0.0 H= 0.0  
 THE CURRENT VALUE OF X IS  
 0.1264958D 01 0.1659086D 01  
 THE CONSTRAINT VALUES  
 0.5896728D-01 0.7595534D-01

1ST ORDER ESTIMATES  
 F= 0.7722596D 01 P= 0.9320005D 01 G= 0.7645402D 01 RSIGNA= 0.0 H= 0.0  
 THE CURRENT VALUE OF X IS  
 0.1243025D 01 0.1651169D 01  
 THE CONSTRAINT VALUES  
 0.106585D 00 0.1058063D 00

LAGRANGE MULTIPLIERS  
 F= 0.7722596D 01 P= 0.9320005D 01 G= 0.7645402D 01 RSIGNA= 0.0 H= 0.0  
 THE CURRENT VALUE OF X IS  
 0.1243025D 01 0.1651169D 01  
 THE CONSTRAINT VALUES  
 0.1567669D 01 0.2319992D 01  
 APPARENTLY ROUND-OFF ERRORS PREVENT A MORE ACCURATE DETERMINATION OF THE MINIMUM OF THIS SUBPROBLEM.  
 TIME= 0.750 SECONDS

14 POINT= 13 DOUT= 0.1185880D-03 PHASE= 2  
 F= 0.7671997D 01 P= 0.8614190D 01 RHO= 0.1562500D 00 MAGNITUDE= 0.1157642D 00 H= 0.0  
 THE CURRENT VALUE OF X IS  
 0.1249429D 01 0.1673076D 01 RSIGNA= 0.7421929D 00 H= 0.0  
 THE CONSTRAINT VALUES  
 0.1128031D 00 0.7669464D-01

2ND ORDER ESTIMATES  
 F= 0.7381120D 01 P= 0.7823229D 01 G= 0.7392012D 01 RSIGNA= 0.0 H= 0.0  
 THE CURRENT VALUE OF X IS  
 0.1300560D 01 0.1693207D 01  
 THE CONSTRAINT VALUES  
 0.1761123D-02 0.6237300D-02

1ST ORDER ESTIMATES  
 F= 0.7432165D 01 P= 0.7916718D 01 G= 0.7392012D 01 RSIGNA= 0.0 H= 0.0  
 THE CURRENT VALUE OF X IS  
 0.1300560D 01 0.1693207D 01

Figure 3.--continued

1. 1473233-02 J-12460360-01

```

LAGRANGE MULTIPLIERS
  P= 0.76221650 01      P= 0.79167780 01      G= 0.7392012D 01      RSIGNA= 0.0      H= 0.0
THE CURRENT VALUE OF X IS
  0.12969673 01
THE CONSTRAINT VALUES
  0.1690579D 01
  0.1385157D 01
  0.2037300D 01
APPARENTLY ROUND-OFF ERRORS PREVENT A MORE ACCURATE DETERMINATION OF THE MINIMUM OF THIS SUBPROBLEM.
TIME= 0.879 SECONDS

```

```

END) ORDER ESTIMATES
F= 0.736131D 01      P= 0.7482046D 01      G= 0.7367544D 01      H= 0.0
THE CURRENT VALUE OF X IS
0.1302893D 01      0.1697103D 01
THE CONSTRAINT VALUES
-0.4280923D-03      0.3936434D-05

```

```

LAST ORDER ESTIMATES
F= 0.7168379D 01      P= 0.7509236D 01      G= 0.7367544D 01      RSIGMA= 0.0      H= 0.0
THE CURRENT VALUE OF X IS
0.1302522D 01      0.1696695D 01
THE CONSTRAINT VALUES
0.1303280D-03      0.7824631D-03
0.1303280D-03

```

```

LAGRANGE MULTIPLIERS
F= 0.7368790 01 P= 0.75092360 01 G= 0.73675440 01 RSIGNA= 0.0 H= 0.0
THE CURRENT VALUE OF X IS
0.13025220 01 0.16966950 01
THE CONSTRAINT VALUES
0.13550630 01 0.19767970 01
APPARENTLY ROUND-OFF ERRORS PREVENT A MORE ACCURATE DETERMINATION OF THE MINIMUM OF THIS SUBPROBLEM.
TIME= 1.003 SECONDS

```

```

2ND ORDER ESTIMATES
F = 0.7366962D 01      P = 0.7395569D 01      G = 0.736696D 01      RSIGMA= 0.0      M= 0.0
THE CURRENT VALUE OF X IS
0.1302774D 01          0.1697236D 01
THE CONSTRAINT VALUES
0.1460523D-03          0.3932242D-04

```

EST ORDER ESTIMATES

Figure 3.--continued



```

THE CURRENT VALUE OF X IS
0.13027173 01 0.16972030 01
THE CONSTRAINT VALUES
0.16507197-03 0.85766720-04

LAGRANGE MULTIPLIERS
F= 0.73670510 01 P= 0.74026730 01 G= 0.73669960 01 R= 0.0
THE CURRENT VALUE OF X IS
0.13027120 01 0.16972030 01
THE CONSTRAINT VALUES
0.13287370 01 0.19513870 01
APPARENTLY ROUND-OFF ERRORS PREVENT A MORE ACCURATE DETERMINATION OF THE MINIMUM OF THIS SUBPROBLEM.
TIME= 1.102 SECONDS

*****
(17) POINT= 19 DOTT= 0.20282430-05 PHASE= 2
F= 0.73715670 01 P= 0.74032310 01 RHO= 0.24414060-02 MAGNITUDE= 0.70276280-01
THE CURRENT VALUE OF X IS G= 0.73666850 01 RSIGMA= 0.31683760-01 M= 0.0
0.13019160 01 0.16968320 01
THE CONSTRAINT VALUES
0.18663760-02 0.12518430-02

2ND ORDER ESTIMATES
F= 0.73666830 01 P= 0.73739130 01 G= 0.73667030 01 R= 0.0
THE CURRENT VALUE OF X IS
0.13027770 01 0.16972280 01
THE CONSTRAINT VALUES
-0.21372170-07 -0.46807200-05

1ST ORDER ESTIMATES
F= 0.73667060 01 P= 0.73757100 01 G= 0.73667030 01 R= 0.0
THE CURRENT VALUE OF X IS
0.13027730 01 0.16972260 01
THE CONSTRAINT VALUES
0.13222670 01 0.19502500 01
APPARENTLY ROUND-OFF ERRORS PREVENT A MORE ACCURATE DETERMINATION OF THE MINIMUM OF THIS SUBPROBLEM.
TIME= 1.219 SECONDS

*****
LAGRANGE MULTIPLIERS
F= 0.73667060 01 P= 0.73757100 01 G= 0.73667030 01 R= 0.0
THE CURRENT VALUE OF X IS
0.13027730 01 0.16972260 01
THE CONSTRAINT VALUES
0.13222670 01 0.19502500 01
APPARENTLY ROUND-OFF ERRORS PREVENT A MORE ACCURATE DETERMINATION OF THE MINIMUM OF THIS SUBPROBLEM.
TIME= 1.219 SECONDS

*****
(18) POINT= 21 DOTT= 0.55388060-06 PHASE= 2
F= 0.73679130 01 P= 0.73775240 01 RHO= 0.61035160-03 MAGNITUDE= 0.80343090-01
THE CURRENT VALUE OF X IS G= 0.73666920 01 RSIGMA= 0.96112440-02 M= 0.0
0.13025600 01 0.16971260 01
THE CONSTRAINT VALUES
0.46235060-03 0.31344850-03

2ND ORDER ESTIMATES
F= 0.73666940 01 P= 0.73684970 01 G= 0.73666940 01 R= 0.0
THE CURRENT VALUE OF X IS
0.13027770 01 0.16972240 01
THE CONSTRAINT VALUES
0.17612570-06 0.62494790-06

```

Figure 3.--continued

[illegible]

```

LAPLACE MULTIPLIERS      P= 0.7368940 01  G= 0.7366940 01  RSIG4= 0.0  M= 0.0
T= CURRENT VALUE OF X IS
0.13027750 01
THE CONSTANT VALUES
0.19472150 01
0.1321003 01
APPROXIMATE MINIMUM ERRORS PREVENT A MORE ACCURATE DETERMINATION OF THE MINIMUM OF THIS SUBPROBLEM.
TIME= 1.371 SECONDS

```

```

(19) .....
PQINT= 23      DUT= 0.4275720 06      PHASE= 2
I= 0.73669930 01      P= 0.73668230 01      MAGNITUDE= 0.20081460 00
THE CURRENT VALUE OF X IS      M= 0.0
0.13027230 01
THE CONSTRAINT VALUES
3.11404160-03      ASIGNA= 0.28304970-02
0.77034420-04      RHO= 0.15256790-03
G= 0.73666880 01

```

```

2ND) ORDER ESTIMATES
F= 0.736686D 01
THE CURRENT VALUE OF X IS
J.130277D 01 0.169722D 01
THE CONSTRAINT VALUES
-0.226618D-05 -0.190320D-05
G= 0.736686D 01 RSIGNA= 0.0 H= 0.0

```

```

1ST ORDER ESTIMATES
F= 0.736686D 01      P= 0.7367256D 01      G= 0.736686D 01      H= 0.0
THE CURRENT VALUE OF X IS
0.130277D 01      0.169725D 01
THE CONSTRAINT VALUES
-0.207303D-05      -0.174360D-05

```

```

LAGRANGE MULTIPLIERS
  F= 0.736696D 01      P= 0.7367256D 01      G= 0.7366696D 01      RSIGNA= 0.0      H= 0.0
THE CURRENT VALUE OF X IS
  0.130277D 01      0.1697225D 01
THE CONSTRAINT VALUES
  0.1330007D 01      0.733261D 01
APPARENTLY ROUND-OFF ERRORS PREVENT A MORE ACCURATE DETERMINATION OF THE MINIMUM OF THIS SUBPROBLEM.
TIME= 1.461 SECONDS

```

```

*****
(20) POINT= 25 DDTT= 0.4806442D-07 PHASE= 2
F= 0.7366763D 01 P= 0.7367581D 01 M= 0.0
THE CURRENT VALUE OF X IS
0.1302762D 01
THE CONSTRAINT VALUES
0.2886958D-04 0.1951167D-04
RHO= 0.3814697D-04 MAGNITUDE= 0.8142487D-01
G= 0.7366692D 01 RSIGMA= 0.8124364D-03

```

```

END OF GROUP ESTIMATES
F= 0.73666400 01
THE CURRENT VALUE OF X IS
0.13027750 01
THE CONSTRAINT VALUES
0.44567685-04 0.46064320-04
G= 0.73666400 01
RSIGNA= 0.0
H= 0.0

```

Figure 3.--continued

1ST ORDER ESTIMATES  
 P= 0.736692D 01 P= 0.736693D 01 G= 0.736694D 01 R= 0.0  
 THE CURRENT VALUE OF X IS  
 0.1302775 01  
 THE CONSTRAINT VALUES  
 0.466193D-06 0.3307564D-06

LAGRANGE MULTIPLIERS  
 F= 0.736694D 01 P= 0.736693D 01 G= 0.736694D 01 R= 0.0  
 THE CURRENT VALUE OF X IS  
 0.1302775 01  
 THE CONSTRAINT VALUES  
 0.1321768 01 0.1955085D 01  
 APPARENTLY ROUND OFF ERRORS PREVENT A MORE ACCURATE DETERMINATION OF THE MINIMUM OF THIS SUBPROBLEM.  
 TIME= 1.621 SECONDS

2ND ORDER ESTIMATES  
 P= 0.736692D 01 P= 0.7366721D 01 G= 0.736692D 01 R= 0.0  
 THE CURRENT VALUE OF X IS  
 0.1302776 01 0.1697224D 01  
 THE CONSTRAINT VALUES  
 -0.1689720D-06 -0.1449764D-06

1ST ORDER ESTIMATES  
 P= 0.736692D 01 P= 0.7366728D 01 G= 0.736692D 01 R= 0.0  
 THE CURRENT VALUE OF X IS  
 0.1302776D 01 0.1697224D 01  
 THE CONSTRAINT VALUES  
 -0.1292761D-06 -0.1152431D-06

LAGRANGE MULTIPLIERS  
 P= 0.736692D 01 P= 0.7366728D 01 G= 0.736692D 01 R= 0.0  
 THE CURRENT VALUE OF X IS  
 0.1302776D 01 0.1697224D 01  
 THE CONSTRAINT VALUES  
 0.1339765D 01 0.1990352D 01  
 APPARENTLY ROUND OFF ERRORS PREVENT A MORE ACCURATE DETERMINATION OF THE MINIMUM OF THIS SUBPROBLEM.  
 TIME= 1.711 SECONDS

2ND ORDER ESTIMATES  
 P= 0.736692D 01 P= 0.7366728D 01 G= 0.736692D 01 R= 0.0  
 THE CURRENT VALUE OF X IS  
 0.1302776D 01 0.1697224D 01  
 THE CONSTRAINT VALUES  
 0.1339765D 01 0.1990352D 01  
 APPARENTLY ROUND OFF ERRORS PREVENT A MORE ACCURATE DETERMINATION OF THE MINIMUM OF THIS SUBPROBLEM.  
 TIME= 1.711 SECONDS

LAGRANGE MULTIPLIERS  
 P= 0.736692D 01 P= 0.7366728D 01 G= 0.736692D 01 R= 0.0  
 THE CURRENT VALUE OF X IS  
 0.1302776D 01 0.1697224D 01  
 THE CONSTRAINT VALUES  
 0.1339765D 01 0.1990352D 01  
 APPARENTLY ROUND OFF ERRORS PREVENT A MORE ACCURATE DETERMINATION OF THE MINIMUM OF THIS SUBPROBLEM.  
 TIME= 1.711 SECONDS

LAGRANGE MULTIPLIERS  
 P= 0.736692D 01 P= 0.7366728D 01 G= 0.736692D 01 R= 0.0  
 THE CURRENT VALUE OF X IS  
 0.1302776D 01 0.1697224D 01  
 THE CONSTRAINT VALUES  
 0.1339765D 01 0.1990352D 01  
 APPARENTLY ROUND OFF ERRORS PREVENT A MORE ACCURATE DETERMINATION OF THE MINIMUM OF THIS SUBPROBLEM.  
 TIME= 1.711 SECONDS

Figure 3.--continued

Figure 3.--continued

0.7366693D-01 0.7366693D-07

1.7 HADJEM PST144T.S  
P= 0.7366693D 01 P= 0.7366701D 01 G= 0.7366693D 01 H= 0.0  
THE CURRENT VALUE OF X IS  
0.13027761 01 0.1697224D 01  
THE CONSTRAINT VALUES  
0.0302841D-07 0.7258831D-07

LAGRANGE MULTIPLIERS  
P= 0.7366693D 01 P= 0.7366701D 01 G= 0.7366693D 01 H= 0.0  
THE CURRENT VALUE OF X IS  
0.13027761 01 0.1697224D 01  
THE CONSTRAINT VALUES  
0.1305098D 01 0.1903826D 01

# SENSITIVITY ANALYSIS

23 THE VALUE OF R(RHO) IS 0.23842D-05

THE POINT AT WHICH THE ESTIMATE OF SENSITIVITY WILL BE MADE IS  
X( 1)= 1.302775 X( 2)= 1.697224 X(

PARAMETER VALUE DIFFERENCING INTERVAL  
1 0.0 0.10000D-09  
2 -3.00000 0.10000D-09

# OPTIMAL VALUE FUNCTION SENSITIVITY

24 DF/DX( 1)= 1.305098 DF/DX( 2)= 1.903824 DF/DX(

25 DETAILED SENSITIVITY RESULTS FOLLOW FOR PARAMETERS  
1, 2,

26 X-DERIVATIVES ARE WITH RESPECT TO PARAMETER , 1  
DX( 1)=0.2773500 DX( 2)= 0.2773500 DX(

27 U-DERIVATIVES WITH RESPECT TO PARAMETER 1  
DX( 1)= 0.5063357 DX( 2)=0.4622335D-01 DX(

28 DFIX(2)/DX= 1.32820  
\*\*\*\*\*

29 X-DERIVATIVES ARE WITH RESPECT TO PARAMETER , 2  
DX( 1)=0.2773497 DX( 2)=0.7226484 DX(

U-DERIVATIVES WITH RESPECT TO PARAMETER 2  
DX( 1)=0.4622325D-01 DX( 2)= 1.308571 DX(

DFIX(1)/DX= 1.93375  
\*\*\*\*\*



```

1.17 DATA ESTIMATES
P= 0.62613650 01    P= 0.62613650 01    RSIGMA= 0.0    H= 0.0
THE CURRENT VALUE OF X IS
0.14384470 01
THE CONSTRAINT VALUES
-0.15333153-06
-0.35286570-06

LAGRANGE MULTIPLIERS
P= 0.62613650 01    P= 0.62613650 01    RSIGMA= 0.0    H= 0.0
THE CURRENT VALUE OF X IS
0.14384470 01
THE CONSTRAINT VALUES
0.20602070 01
0.91699533 00
APPARENTLY ROUNDING OFF ERRORS PREVENT A MORE ACCURATE DETERMINATION OF THE MINIMUM OF THIS SUBPROBLEM.
TIME= 2.391 SECONDS

*****
POINT= 23    DOTT= 0.90250670-07
P= 0.62613650 01    P= 0.62613660 01    RSIGMA= 0.0    H= 0.0
THE CURRENT VALUE OF X IS
0.14384470 01
THE CONSTRAINT VALUES
0.15615490 01
0.14384460 01
THE CONSTRAINT VALUES
0.10230960-04    0.65429340-05
PHASE= 2

2ND ORDER ESTIMATES
P= 0.62613660 01    P= 0.62613940 01    RSIGMA= 0.0    H= 0.0
THE CURRENT VALUE OF X IS
0.15615530 01
THE CONSTRAINT VALUES
0.14384470 01
-0.1906200-04    -0.98911230-07

1ST ORDER ESTIMATES
P= 0.62613660 01    P= 0.62614020 01    RSIGMA= 0.0    H= 0.0
THE CURRENT VALUE OF X IS
0.15615530 01
THE CONSTRAINT VALUES
0.14384470 01
-0.22557890-06    -0.11478340-06

LAGRANGE MULTIPLIERS
P= 0.62613660 01    P= 0.62614020 01    RSIGMA= 0.0    H= 0.0
THE CURRENT VALUE OF X IS
0.15615530 01
THE CONSTRAINT VALUES
0.14384470 01
0.93215540 00    0.20992480 01
APPARENTLY ROUNDING OFF ERRORS PREVENT A MORE ACCURATE DETERMINATION OF THE MINIMUM OF THIS SUBPROBLEM.
TIME= 2.500 SECONDS

*****
POINT= 25    DOTT= 0.29651740-07
P= 0.62613710 01    P= 0.62614340 01    RSIGMA= 0.0    H= 0.0
THE CURRENT VALUE OF X IS
0.15615520 01
THE CONSTRAINT VALUES
0.14384470 01
0.26639090-05
PHASE= 2

2ND ORDER ESTIMATES
P= 0.62613670 01    P= 0.62613730 01    RSIGMA= 0.0    H= 0.0
THE CURRENT VALUE OF X IS
0.15615530 01
THE CONSTRAINT VALUES
0.14384470 01
0.14384470 01

```

Figure 3.--continued

THE CURRENT ESTIMATE VALUES  
 F= 0.15501367D-06 P= 0.6261375D-01

1ST ORDER ESTIMATES  
 F= 0.6261367D-01 P= 0.6261375D-01 G= 0.6261367D-01 H= 0.0  
 THE CURRENT VALUE OF X IS  
 0.1561553D-01 0.1438867D-01  
 THE CONSTRAINT VALUES  
 0.161553D-06 0.7172925D-07

LAGRANGE MULTIPLIERS  
 F= 0.6261367D-01 P= 0.6261375D-01 G= 0.6261367D-01 H= 0.0  
 THE CURRENT VALUE OF X IS  
 0.1561553D-01 0.1438867D-01  
 THE CONSTRAINT VALUES  
 0.8949952D-06 0.2004308D-01

### SENSITIVITY ANALYSIS

THE VALUE OF R(RHO) IS 0.23842D-05

THE POINT AT WHICH THE ESTIMATE OF SENSITIVITY WILL BE MADE IS  
 X(1)= 1.561552 X(2)= 1.438867 X(

PARAMETER VALUE DIFFERENCING INTERVAL  
 1 -1.00000 0.10000D-09  
 2 -3.00000 0.10000D-09

### OPTIMAL VALUE FUNCTION SENSITIVITY

33 DF/DA(1)= 0.8949943 DF/DA(2)= 2.004307 DF/DA(

Figure 3.--continued

OPTIMAL VALUE FUNCTION BOUNDS AND F\* PAR(1) IS PERTINENT  
 \*\*\*\*\*

POINT 1 (UNPERTURBED SOLUTION) : F(PAR(1)) = 0.73666930 01  
 PAR(1) = 0.2316630-15

(34)

POINT 2 (PERTURBED SOLUTION) : F(PAR(1)) = 0.62613670 01  
 PAR(1) = -0.10000000 01

LINE PASSING THROUGH POINTS 1 AND 2 AND OVER ESTIMATING F\*

(35)

F = 0.11053260 01 \* PAR(1) + 0.73666930 01

LINE UNDER ESTIMATING F\* AT POINT 1

(36)

F = 0.13350990 01 \* PAR(1) + 0.73666930 01

LINE UNDER ESTIMATING F\* AT POINT 2

(37)

F = 0.89499430 00 \* PAR(1) + 0.71563610 01

QUADRATIC ESTIMATION OF F\* THROUGH POINTS 1 AND 2

(38)

F = 0.19977190 00 \* PAR(1)\*\*2 + 0.13050980 01 \* PAR(1) + 0.73666930 01

F\* BOUND EVALUATION AT TEN EQUIDISTANCE POINTS BETWEEN POINTS 1 AND 2

PAR(1)	LOWER BOUND	UPPER BOUND	QUAD ESTIMATE
0.000	0.73666930 01	0.73666930 01	0.73666930 01
-0.100	0.72381830 01	0.72561600 01	0.72381800 01
-0.200	0.71056730 01	0.71496270 01	0.71136640 01
-0.300	0.69751630 01	0.70350950 01	0.69931430 01
-0.400	0.68446530 01	0.69245620 01	0.68766170 01
-0.500	0.67141440 01	0.68140300 01	0.67640870 01
-0.600	0.65836340 01	0.67034970 01	0.66555520 01
-0.700	0.64531240 01	0.65929640 01	0.65510120 01
-0.800	0.63226140 01	0.64824320 01	0.64504680 01
-0.900	0.61921040 01	0.63718990 01	0.63539200 01
-1.000	0.60615940 01	0.62613670 01	0.62613670 01

(39)

Figure 3.--continued



IV

MULTIPLIER PROGRAMMING ROUTINE-SUMT VERSION 4 03/01/73

```

1= 2 4= 2 42= 0
NAT. TIME= 0.900000E 03 R= 0.100000E 02 RATIO= 0.400000E 01 EPSILON= 0.9999997E-08 THETA= 0.9999994E-06
OPTIMIS SELECTED 1 1 1 1 1 1 0 1 1 1
TOLERANCES
0.0 0.9999999E-03
SECOND SET OF OPTIONS
1 1 1 1 1 0 1
TIME= 0.012 SECONDS
F= 0.9250300 01 P= 0.0
THE CURRENT VALUE OF X IS
0.1000000 01 0.1500000 01
THE CONSTRAINT VALUES
0.5000000 00 -0.1500000 01
APPARENTLY ROUND-OFF ERRORS PREVENT A MORE ACCURATE DETERMINATION OF THE MINIMUM OF THIS SUBPROBLEM.
TIME= 0.219 SECONDS
*****
POINT= 3 DOTT= 0.1492879D-03 RHO= 0.1000000 02 MAGNITUDE= 0.1215422D-01 PHASE= 1
F= 0.8750897D 01 P= -0.1427585D 02 G= -0.1124910D 02 RSIGMA= -0.2302675D 02 H= 0.0
THE CURRENT VALUE OF X IS
-0.5003024D 00 0.1025120D 02
THE CONSTRAINT VALUES
0.9999103D 00 -0.1142740D 01
TIME= 0.179 SECONDS
*****
LAGRANGE MULTIPLIERS
F= 0.8750897D 01 P= -0.1427585D 02 G= -0.1124910D 02 RSIGMA= -0.2302675D 02 H= 0.0
THE CURRENT VALUE OF X IS
-0.5003024D 00 0.1025120D 02
THE CONSTRAINT VALUES
0.9999103D 00 -0.1142740D 01
TIME= 0.179 SECONDS
*****
THE FEASIBLE STARTING POINT TO BE USED IS ...
F= 0.2050444D 02 P= 0.1279110D 02 G= 0.0
THE CURRENT VALUE OF X IS
-0.4999494D 00 0.1495133D 01
THE CONSTRAINT VALUES
0.1245184D 01 0.4916430D-02
APPARENTLY ROUND-OFF ERRORS PREVENT A MORE ACCURATE DETERMINATION OF THE MINIMUM OF THIS SUBPROBLEM.
TIME= 0.060 SECONDS
*****
POINT= 7 DOTT= 0.8961926D-05 RHO= 0.1000000 02 MAGNITUDE= 0.1689025D-01 PHASE= 2
F= 0.2038678D 02 P= 0.3018354D 02 G= 0.4677850D-01 RSIGMA= 0.1009676D 02 H= 0.0
THE CURRENT VALUE OF X IS
-0.4016757D 00 0.7420775D 00
THE CONSTRAINT VALUES
0.6510695D 00 0.5595977D 00
*****
LAGRANGE MULTIPLIERS
F= 0.2038678D 02 P= 0.3018354D 02 G= 0.4677850D-01 RSIGMA= 0.1009676D 02 H= 0.0
THE CURRENT VALUE OF X IS
-0.4016757D 00 0.7420775D 00
THE CONSTRAINT VALUES
0.6510695D 00 0.5595977D 00

```

Figure 3.--continued

OPTIMAL VALUE FUNCTION ROUNDS WHEN PART 2) IS PERTURBED  
\*\*\*\*\*

POINT 1 (UNPERTURBED SOLUTION) : FIPART 2) = 0.73666930 01  
PART 2) = -0.30000000 01  
POINT 2 (PERTURBED SOLUTION) : FIPART 2) = 0.14055730 02  
PART 2) = -0.10000000 01

LINE PASSING THROUGH POINTS 1 AND 2 AND OVER ESTIMATING F\*

F = 0.33445180 01 \* PART 2) + 0.17400250 02

LINE UNDER ESTIMATING F\* AT POINT 1

F = 0.19030240 01 \* PART 2) + 0.13078170 02

LINE UNDER ESTIMATING F\* AT POINT 2

F = 0.47961840 01 \* PART 2) + 0.18851910 02

QUADRATIC ESTIMATION OF F\* THROUGH POINTS 1 AND 2

F = 0.72034670 00 \* PART 2) + 0.62259050 01 \* PART 2) + 0.19561290 02

F\* ROUND EVALUATION AT TEN EQUIDISTANCE POINTS BETWEEN POINTS 1 AND 2

PART 2)	LOWER BOUND	UPPER BOUND	QUAD ESTIMATE
-2.000	0.73666930 01	0.73666930 01	0.73666930 01
-1.800	0.77474570 01	0.90355960 01	0.77762710 01
-1.600	0.81282220 01	0.87015000 01	0.82434780 01
-1.400	0.85089870 01	0.93734030 01	0.87683120 01
-1.200	0.88897520 01	0.10042310 02	0.93507740 01
-1.000	0.92705170 01	0.10711210 02	0.99908640 01
-0.800	0.10218780 02	0.11369110 02	0.10688540 02
-0.600	0.11178020 02	0.12049020 02	0.11443930 02
-0.400	0.12137250 02	0.12717920 02	0.12256900 02
-0.200	0.13096490 02	0.13386820 02	0.13127500 02
-0.000	0.14055730 02	0.14055730 02	0.14055730 02

Figure 3.---continued

I Output for solution of the unperturbed problem, i.e.,  $\epsilon_1 = 0$ ,  $\epsilon_2 = -3$  (Step 1 of Algorithm 3.1).

- ① Printout of parameter card data.
- ② Printout of first option card data.
- ③ Printout of tolerance card data.
- ④ Printout of second option card data.
- ⑤ Elapsed time since the start of the problem. (This information is not accurate. Subroutine TIMEC, which monitors the elapsed time, has to be modified.)
- ⑥ Initial starting point  $x^0$  and corresponding problem function values.
- ⑦ Feasibility of  $x^0$  is verified. Note that if  $x^0$  is not feasible, then SENSUMT invokes the subroutines FEAS and BODY to find a feasible starting point.
- ⑧ The solution to the first subproblem. Program took four inverse product moves to minimize  $W$  for

$$RH0 = 10, (r=10)$$

$$D0TT = \nabla_x W[x(r)]^T \left[ \frac{\partial^2 W[x(r)]}{\partial x_i \partial x_j} \right]^{-1} \nabla_x W[x(r)] < 10^{-7}$$

$$MAGNITUDE = ||\nabla_x W[x(r)]|| \quad (\text{the magnitude of the gradient of } W)$$

$$F = f[x(r)] \quad (\text{the value of the objective function})$$

$$W = f[x(r)] - r \sum_{j=1}^{\infty} \ln g_j[x(r)] + \sum_{j=M+1}^{M+MZ} [h_j[x(r)]]^2 / r$$

(the value of the P-function)

Note: The  $W$ -function coded in SENSUMT differs from that given in (2.6) in the last term by a factor of 2.

$$\text{RESIGMA} = -r \sum_{j=1}^M \ln g_j[x(r)] \quad (\text{note that this may be negative})$$

$$H = \sum_{j=M+1}^{M+MZ} h_j^2[x(r)]/r$$

$$G = \text{dual value} = f[x(r)] - r^*M + 2.*H$$

(in a convex problem  $F \geq v^* \geq G$ )

where the current value of  $x$  is  $x(r)$  and the current constraint values are  $g_1[x(r)]$  and  $g_2[x(r)]$ .

- ⑨ Values of  $F$ ,  $W$  (designated by  $P$  in the computer output),  $G$ ,  $\text{RSIGMA}$ , and  $x$  are repeated. Current constraint values (i.e.,  $u_j$  and  $w_j$ ) are  $r/g_j$ ,  $j=1, M$ , and  $2n_j(x)/r$ ,  $j=M+1, M+MZ$ . Note: Since Example 1 has no equality constraint, the entries relating to  $h_j$  in 8 and 9 are zero.
- ⑩ The solution to the second subproblem ( $r = 10/4 = 2.5$ ).
- ⑪ The current first order estimates of  $x$  (obtained by first order extrapolation, using the solution of the first and second subproblems) and the corresponding problem functions.
- ⑫ The solution to the third subproblem ( $r = 2.5/4 = .625$ ).
- ⑬ The current second order estimates of  $x$  (obtained by second order extrapolation using solutions of the first, second, and third subproblems) and the corresponding problem functions.
- ⑭ The solution to the fourth subproblem.
- ⑮ The solution to the fifth subproblem.
- ⑯ The solution to the sixth subproblem.
- ⑰ The solution to the seventh subproblem.
- ⑱ The solution to the eighth subproblem.
- ⑲ The solution to the ninth subproblem.
- ⑳ The solution to the tenth subproblem.

- ②① The solution to the eleventh subproblem.
  - ②② The solution to the twelvth (final) subproblem.
  - ②③ The general information about the sensitivity analysis including the value of  $r$ , the estimated final solution point and the parameter values and associated differencing intervals.
  - ②④ Estimates of the sensitivity of the optimal value function obtained by taking the gradient of the Lagrangian with respect to the parameters as described in Step 9 of Algorithm 2.1.
  - ②⑤ The parameters for which detailed sensitivity results follow. Here it is both parameters since the optimal value function is sensitive to both.
  - ②⑥ Estimates of the partial derivatives of the solution point taken with respect to parameter one as in Steps 1-3 of Algorithm 2.1.
  - ②⑦ Estimates of the partial derivatives of the Lagrange multipliers taken with respect to parameter one as in Step 5 of Algorithm 2.1.
  - ②⑧ Estimate of the partial derivative of the optimal value function taken with respect to parameter one and obtained by using the chain rule as in Step 8 of Algorithm 2.1.
  - ②⑨ Same as ②⑥, ②⑦, and ②⑧ but with respect to parameter two.
- II      Output for solution of the first perturbed problem, i.e.,  $\epsilon_1 = -1$ ,  $\epsilon_2 = -3$  (Step 2 of Algorithm 3.1).
- ③① Printout of the input data and subproblem solutions for the first perturbed problem.
  - ③② Printout of the final subproblem for the first perturbed problem.
  - ③③ Same as ②③ but for the first perturbed problem.
  - ③④ Same as ②④ but for the first perburbed problem.

## III

Output for optimal value function bounds as a function of first parameter (Steps 3 and 4 of Algorithm 3.1).

- (34)  $(f^*(\epsilon_1), \epsilon_1)$  of the unperturbed and first perturbed problems [see (22), (23), and (31), (32), respectively].
- (35) Equations of parametric upper bound on  $f^*(\epsilon_1)$ .
- (36) Equations of parametric lower bounds on  $f^*(\epsilon_1)$ .
- (37) Maximum of these equations over the range of  $-1 \leq \epsilon_1 \leq 0$  provides a lower bound on  $f^*(\epsilon_1)$ .
- (38) Equation of a quadratic estimate of  $f^*(\epsilon_1)$  [see the description of subroutine BOUND in Section 6 for the method of deriving this equation].
- (39) Value of bounds and the quadratic estimate of  $f^*(\epsilon_1)$  at eleven equidistant points over the perturbation range of the first parameter  $\epsilon_1$ .

## IV

Output for solution of the second perturbed problem, i.e.,  $\epsilon_1 = 0$ ,  $\epsilon_2 = -1$  (Step 2 of Algorithm 3.1).

Description of the output for the second perturbed problem is similar to that of the first perturbed problem, i.e., (30) - (33).

## V

Output for optimal value function bounds as a function of the second parameter (Steps 3 and 4 of Algorithm 3.1).

The description here again closely parallels that of the first parameter, i.e., (34) - (39).

The graphical depiction of the bounds derived for  $f^*(\epsilon)$  as a function of  $\epsilon_2$ , together with a plot of the analytical solution and quadratic estimates of  $f^*(\epsilon)$  at different values of  $\epsilon_2$  is in Figure

4.

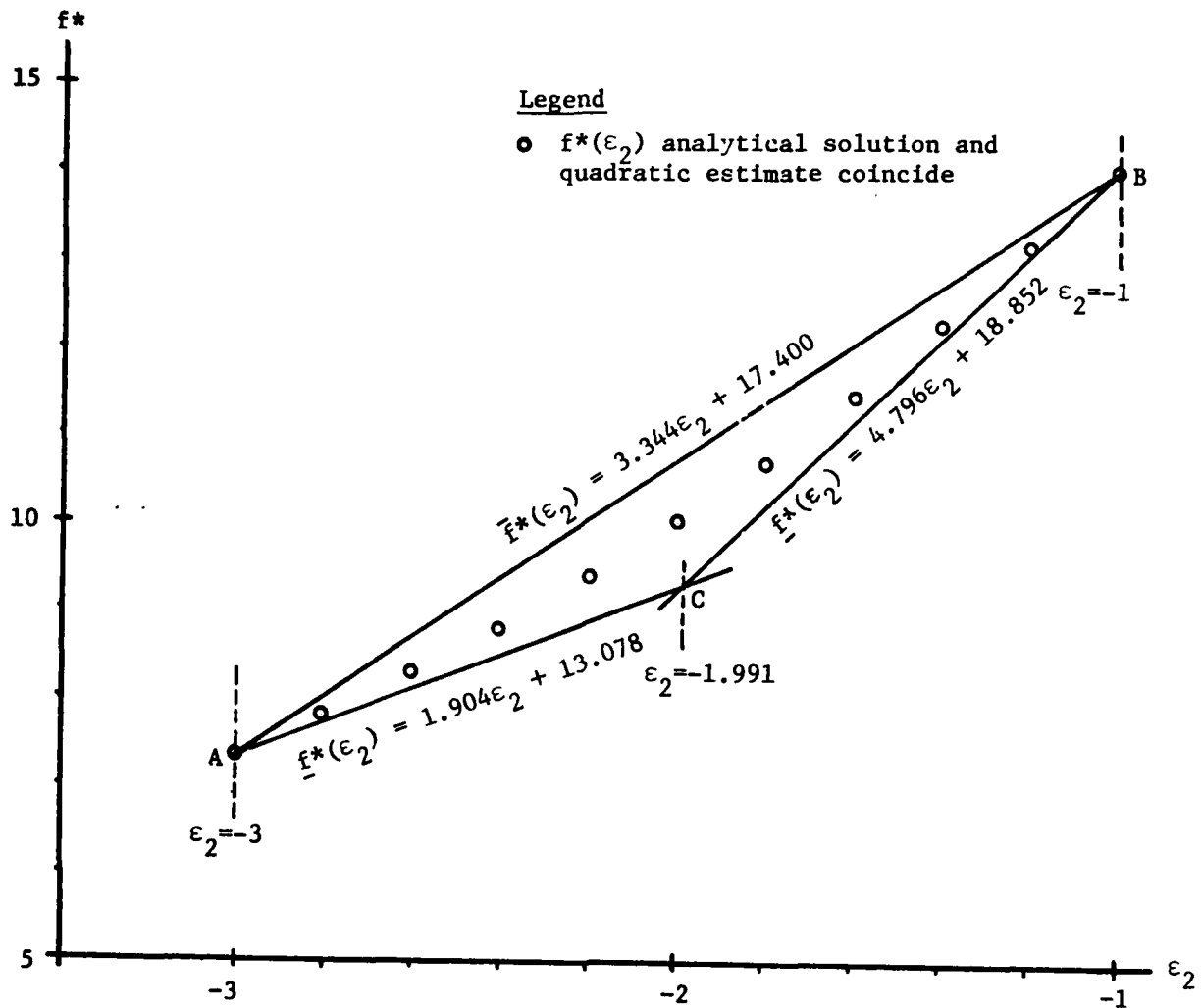


Figure 4.--Graph of bounds on  $f^*(\epsilon_2)$  of Example 1 (computer solution).

## 5.2 Example 2

$$\text{Minimize } f(x) = x_1^2 + .5e^{x_2} - .125x_3^3 + .25(x_4 - 40)^2 - .1e^{x_5} \\ \text{subject to } x_1, \dots, x_5$$

$$\text{Subject to } g_1(x) = -.5x_1^2 + 6x_2 - 5e^{x_3} - .05x_4^2 - .5/x_5 \geq \epsilon_1$$

$$g_2(x) = -5e^{-x_1} - 2x_2^2 + 3x_3 + x_4 + 3x_5 \geq -12$$

$$g_3(x) = 3x_1 + x_2 - x_3^2 + x_4 - x_5^2 \geq 2$$

$$0 \leq x_3 \leq 5, \quad 0 \leq x_5 \leq 5, \quad 0 \leq x_j \leq \infty, \quad j=1,2,4.$$

It is desired to calculate bounds on  $f^*(\epsilon_1)$  of this problem for  $\epsilon_1 \in [-10, 5]$ .

(i) Problem  $SR(\epsilon)$ .

• Standard format:

$$\begin{aligned} \text{Min}_{x_1, \dots, x_5} \quad f(x) = \sum_{j=1}^5 f_j(x_j) = & x_1^2 + .5e^{x_2} - .125x_3^3 \\ & + .25(x_4 - 40)^2 - .1e^{x_5} \end{aligned}$$

S.t.  $x \in G \cap B$ , where

$$G = \{x : g_1(x) \geq \epsilon_1, g_2(x) \geq -12, g_3(x) \geq 2\}, \quad -10 \leq \epsilon_1 \leq 5$$

and

$$B = \{0 \leq x_3 \leq 5, 0 \leq x_5 \leq 5, 0 \leq x_i \leq \infty, i=1,2,4\}.$$

(ii) Problem  $\tilde{CSR}(\epsilon)$ , (convex overestimating problem of  $SR(\epsilon)$ ).

• Problem formulation:

$$\tilde{f}_3(x_3) = -3.125x_3 + 6.014$$

$$\tilde{f}_5(x_5) = -2.948x_5 + 7.027$$

Remaining terms are identical to those of problem  $SR(\epsilon)$ ; thus problem  $\tilde{CSR}(\epsilon)$  reads

$$\begin{aligned} \text{Min}_{x_1, \dots, x_5} \quad \tilde{f}(x) = & x_1^2 + .5e^{x_2} - 3.125x_3 + .25(x_4 - 40)^2 \\ & - 2.948x_5 + 13.041 \end{aligned}$$

S.t.  $x \in \tilde{G} \cap B, \quad -10 \leq \epsilon_1 \leq 5,$

where  $\tilde{G} \supseteq G$ .

Note:  $\tilde{f}_3(x_3)$  and  $\tilde{f}_5(x_5)$  were chosen here to be the lowest nonunderestimating lines parallel to the convex envelopes of  $f_3(x_3)$  and  $f_5(x_5)$ , respectively.



(iii) Problem  $\tilde{CSR}(\epsilon)$ , (convex underestimating problem).

• Problem formulation:

$$\tilde{f}_3(x_3) = -3.125x_3$$

$$\tilde{f}_5(x_5) = -2.948x_5$$

Remaining terms are identical to those of problem  $SR(\epsilon)$ ;  
thus problem  $\tilde{CSR}(\epsilon)$  reads

$$\begin{aligned} \text{Min}_{x_1, \dots, x_5} \quad f(x) = & x_1^2 + .5e^{x_2} - 3.12x_3 + .25(x_4 - 40)^2 \\ & - 2.948x_5 - .1 \end{aligned}$$

$$\text{S.t.} \quad x \in \tilde{G} \cap B, \quad -10 \leq \epsilon_1 \leq 5$$

where  $\tilde{G} \equiv G$ .

The computer listing of the code for problem  $\tilde{CSR}(\epsilon)$  is shown in Figure 5. The listing of the code for  $\tilde{CSR}(\epsilon)$  is identical to that of the problem  $\tilde{CSR}(\epsilon)$  except for the constant term in the objective function, which is + 13.041, rather than - .1.

(iv) Bounds.

Figures 6 and 7 depict the calculated upper and lower bounds via the analysis of the problems  $\tilde{CSR}(\epsilon)$  and  $\tilde{CSR}(\epsilon)$ , respectively. These results are summarized below.

Upper bound:

$$\bar{f}^*(\epsilon_1) = 4.825\epsilon_1 + 136.628.$$

Lower bound:

$$\underline{f}^*(\epsilon_1) = \max[1.880\epsilon_1 + 94.032, 5.466\epsilon_1 + 120.284].$$

The graphical depiction of these bounds is in Figure 8.

```

J=GRAEMI T=2 L=2
// EXEC FORG1
SUBROUTINE READIN
  IMPLICIT REAL*8(A-H,O-Z)
  COMMON/SEN/PAR(45),DPAR(45),NPAR,ISENS
  COMMON/ABG/LY,LZ,PER(45)
  NPAR=3
  PAR(1)=5.
  PAR(2)=-12.
  PAR(3)=2.
  PER(1)=-15.
  RETURN
END
SUBROUTINE RESTINT(IN,VAL)
  IMPLICIT REAL*8(A-H,O-Z)
  COMMON/SHARE/X(45),DEL(45),A(45,45),N,M,MN,NPI,NM1
  COMMON/SEN/PAR(45),DPAR(45),NPAR,ISENS
  COMMON/ABG/LY,LZ,PER(45)
  I=IN+1
  GO TO (20,1,2,3,4,5),I
20 VAL=X(1)**2+.5*DEXP(X(2))-3.125*X(3)+.25*(X(4)-40.)**2
  * -2.948*X(5)+13.041
  RETURN
1 VAL=-.5*X(1)**2+6.*X(2)-5.*DEXP(X(3))-0.5*X(4)**2-.5/X(5)
  * -PAR(1)
  RETURN
2 VAL=-5.*DEXP(-X(1))-2.*X(2)**2+3.*X(3)+X(4)+3.*X(5)-PAR(2)
  RETURN
3 VAL=3.*X(1)+X(2)-X(3)**2+X(4)-X(5)**2-PAR(3)
  RETURN
4 VAL=-X(3)+5.
  RETURN
5 VAL=-X(5)+5.
  RETURN
END
SUBROUTINE GRAD1(IN)
  IMPLICIT REAL*8(A-H,O-Z)
  COMMON/SHARE/X(45),DEL(45),A(45,45),N,M,MN,NPI,NM1
  COMMON/SEN/PAR(45),DPAR(45),NPAR,ISENS
  J=IN+1
  DO 15 I=1,N
    DEL(I)=0.0
15 CONTINUE
  GO TO (20,1,2,3,4,5),J
20 DEL(1)=2.*X(1)
  DEL(2)=.5*DEXP(X(2))
  DEL(3)=-3.125
  DEL(4)=.5*(X(4)-40.)
  DEL(5)=-2.948
  RETURN
1 DEL(1)=-1.*X(1)
  DEL(2)=6.
  DEL(3)=-5.*DEXP(X(3))
  DEL(4)=-.1*X(4)
  DEL(5)=.5/X(5)**2
  RETURN
2 DEL(1)=5.*DEXP(-X(1))
  DEL(2)=-4.*X(2)
  DEL(3)=3.
  DEL(4)=1.
  DEL(5)=3.

```

Figure 5.--Computer listing of the code for the convex overestimating problem of Example 2.

```

      RETURN
3   DEL(1)=3.
      DEL(2)=1.
      DEL(3)=-2.*X(3)
      DEL(4)=1.
      DEL(5)=-2.*X(5)
      RETURN
4   DEL(3)=-1.
      RETURN
5   DEL(5)=-1.
      RETURN
      END
      SUBROUTINE MATRIX(IN,K)
      IMPLICIT REAL*8(A-H,O-Z)
      COMMON/SHARE/X(45),DEL(45), A(45,45),N,M,MN,NPI,NM1
      COMMON/SEN/PAR(45),DPAR(45),NPAR,ISENS
      L=IN+1
      GO TO (20,1,2,3,4,4),L
20  A(1,1)=2.
      A(2,2)=.5*DEXP(X(2))
      A(4,4)=.5
      RETURN
1   A(1,1)=-1.
      A(3,3)=-5.*DEXP(X(3))
      A(4,4)=-.1
      A(5,5)=-1./X(5)**3
      RETURN
2   A(1,1)=-5.*DEXP(-X(1))
      A(2,2)=-4.
      RETURN
3   A(3,3)=-2.
      A(5,5)=-2.
      RETURN
4   K=1
      RETURN
      END
// EXEC FORG6,DSN='0R799661.BOUNDS45',PROG=MAIN
//GO.SYSLIB DD
//          DD
// DD DSN=GNU.FORTLIB,DISP=SHR
//GO.SYSLIN DD
// DD DSN=STEMPUNCH,DISP=(OLD,DELETE)
//F107F001 DD SYSOUT=A
      1.0E-08      10.0      1.0E-06      4.0      900.0      5      5      0
      3.          3.          4.          1.          4.          1          1
      3      1      1      1      1      1
      0.1E-08      0.0001
      1      4      1      1      1      0      2
//

```

Figure 5.--continued

OPTIMAL VALUE FUNCTION BOUNDS WHEN  $\text{PAR}(1)$  IS PERTURBED  
 \*\*\*\*\*

POINT 1 (UNPERTURBED SOLUTION) :  
 $\text{PAR}(1) = 0.5000000 \text{ 01}$   $F(\text{PAR}(1)) = 0.16075470 \text{ 03}$

POINT 2 (PERTURBED SOLUTION) :  
 $\text{PAR}(1) = -0.1000000 \text{ 02}$   $F(\text{PAR}(1)) = 0.98373770 \text{ 02}$

LINE PASSING THROUGH POINTS 1 AND 2 AND OVER ESTIMATING  $f^*$   
 -----

$$f = 0.8253980 \text{ 01} * \text{PAR}(1) + 0.13662770 \text{ 03}$$

FUNCTION EVALUATION AT TEN EQUIDISTANCE POINTS BETWEEN POINTS 1 AND 2  
 =====

$\text{PAR}(1)$	LOWER BOUND	UPPER BOUND	QUAD ESTIMATE
5.000		0.16075470 03	
3.500		0.15351660 03	
2.000		0.14627850 03	
0.500		0.13904040 03	
-1.000		0.13180230 03	
-2.500		0.12456430 03	
-4.000		0.11732620 03	
-5.500		0.11008810 03	
-7.000		0.10285000 03	
-8.500		0.95611870 02	
-10.000		0.98373770 02	

Figure 6.--Parametric upper bound on  $f^*(\epsilon_1)$  via problem  $\tilde{\text{CSR}}(\epsilon)$ ,  
 Example 2 (computer solution).

OPTIMAL VALUE FUNCTION BOUNDS WHEN  $\text{PAR}(1)$  IS PERTURBED  
 \*\*\*\*\*

POINT 1 (UNPERTURBED SOLUTION) :  
 $\text{PAR}(1) = 0.50000000 \text{ 01}$        $f(\text{PAR}(1)) = 0.14761370 \text{ 03}$

POINT 2 (PERTURBED SOLUTION) :  
 $\text{PAR}(1) = -0.10000000 \text{ 02}$        $f(\text{PAR}(1)) = 0.75232770 \text{ 02}$

LINE UNDER ESTIMATING  $f^*$  AT POINT 1  
 -----

$$f = 0.34659210 \text{ 01} * \text{PAR}(1) + 0.12028410 \text{ 03}$$

LINE UNDER ESTIMATING  $f^*$  AT POINT 2  
 -----

$$f = 0.18799810 \text{ 01} * \text{PAR}(1) + 0.94032580 \text{ 02}$$

$f^*$  (QUAD) EVALUATION AT TEN EQUIDISTANCE POINTS BETWEEN POINTS 1 AND 2  
 =====

$\text{PAR}(1)$	LOWER BOUND	UPPER BOUND	QUAD ESTIMATE
5.000	0.14761370 03		
4.500	0.13941490 03		
4.000	0.13121600 03		
3.500	0.12301710 03		
-1.000	0.11481820 03		
-2.500	0.10661930 03		
-4.000	0.99420460 02		
-5.500	0.98221580 02		
-7.000	0.82022700 02		
-8.500	0.78052740 02		
-10.000	0.75232770 02		

Figure 7.--Parametric lower bound on  $f^*(\epsilon_1)$  via problem  $\text{CSR}(\epsilon)$ ,  
 Example 2 (computer solution).

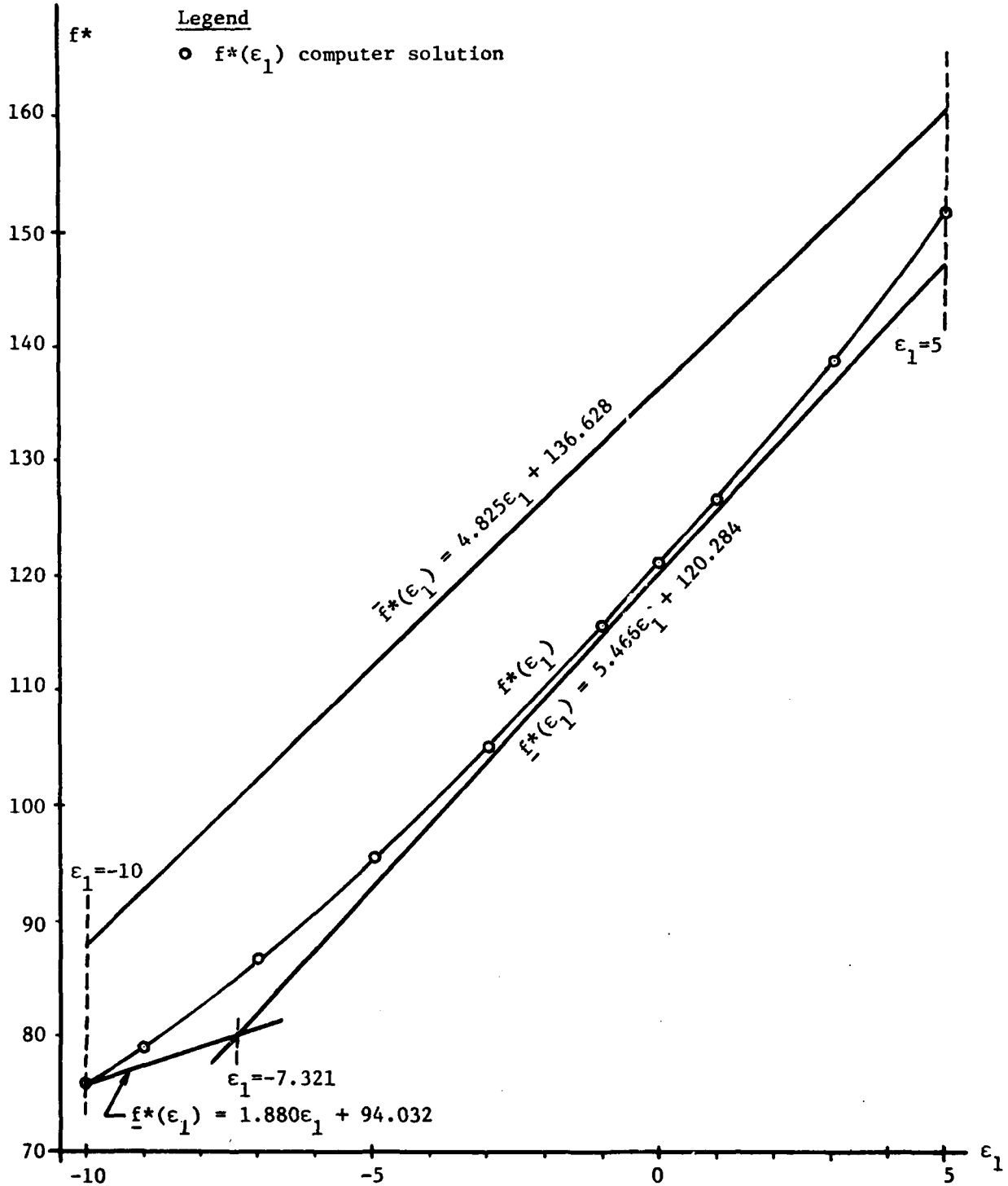


Figure 8.--Parametric bounds on  $f^*(\epsilon)$ , Example 2 (computer solution).

## 6. General Description and Listing of the SENSUMT Subroutines

The subroutines comprising SENSUMT fall into four categories:

- (i) user subroutines,
- (ii) SUMT subroutines,
- (iii) sensitivity subroutines, and
- (iv) bound subroutines.

As mentioned before, user subroutines are generally composed of four subroutines, i.e., READIN, RSTNT, GRAD1, and MATRIX. These subroutines respectively provide pertinent information about the problem, functions of the problem, their gradients and Hessian matrices. A detailed description of these subroutines, together with the corresponding codes for Examples 1 and 2, were given in Section 4 and Figures 2 and 5.

SUMT subroutines implement the Sequential Unconstrained Minimization Technique of Fiacco and McCormick [10]. The subroutines, along with program MAIN, comprising this group are:

• BØDY	• ØUTPUT
• CHCKER	• PEVALU
• CØNVRG	• PUNCH
• DIFF1	• REJECT
• DIFF2	• RHØCØM
• ESTIM	• SECØND
• EVALU	• SECØRD
• FEAS	• SET
• FINAL	• STØRE
• GRAD	• TCHECK
• INVERS	• TIMEC
• MAIN	• TRANS
• ØPT	• XMØVE

With the exception of the subroutine TRANS, the above routines have been developed by Mylander, Holmes, and McCormick [13]. Some of

these routines, in particular program MAIN and subroutines BODY and OUTPUT, have been modified for implementing sensitivity and bound calculation routines. Subroutine TRANS was recently developed to aid the user when analyzing a convex equivalent of geometric programming problems by SENSUMT. See page 104 for a more detailed description of this subroutine.

Sensitivity subroutines implement the sensitivity analysis Algorithm 2.1 and interface it with the SUMT subroutines. A brief history of the development of these subroutines was given in Section 1. The latest version of the codes comprising this group, due to Armacost [1], are subroutines SENS, PARDIF, LMULT, and PRESEN. Subroutines SENS and PRESEN have been slightly modified in implementing the bound calculation routines.

Bound subroutines, developed in [11], implement the bound calculation Algorithms 3.1 and 3.2 given in Section 3. The two subroutines in this group are BOUND and PERT.

All of the subroutines comprising SENSUMT are dimensioned to solve problems having at most 45 variables, 45 parameters, and 200 constraints. However, if the computer capacity permits, they may readily be re-dimensioned to solve problems of larger size. All of these subroutines are separately filed in the Conversational Monitor System (CMS) component of the IBM Virtual Machine facility 37C (VM/370) at The George Washington University Center for Academic and Administrative Computing under their corresponding names.

To make this manual self-contained, the computer listing and a general description of each subroutine is provided (in alphabetical order by name). The descriptions of the SUMT subroutines are largely taken from [13], and those of the sensitivity subroutines are taken from [1]. The user subroutines are problem-dependent, thus they are included in the following listings. For familiarity with the user subroutines, the reader should refer to the coding of Examples 1 and 2 provided in Figures 2 and 5, respectively.



### 6.1 BODY

Subroutine BODY coordinates the flow among the subroutines that actually do the calculations required by the various phases of the algorithm. The flow in this routine is slightly different when the program is in the feasibility phase (solving the entry problem) rather than the normal phase (solving the stated NLP problem). The listing appears as Figure 9.

### 6.2 BOUND

The subroutine BOUND, called by the program MAIN, generates the equations for the upper and lower bounds of the optimal value function as functions of the problem parameters. When analyzing a problem which is known to have a convex or concave optimal value function, BOUND also derives the equation for a quadratic function which approximates the optimal value function. If  $f^*(\epsilon)$  is convex (concave), it corresponds to the higher (lower) of the two quadratic functions, in the perturbation range of the parameter of interest, which passes through the two calculated points in  $f^*(\epsilon), \epsilon$  space and has the calculated slope at each of these points. Evaluation and printout of this quadratic function (when applicable), and the upper and/or lower bound at eleven equidistant points over the perturbation range of the parameters of interest are also programmed in subroutine BOUND. The listing appears as Figure 10.

### 6.3 CHCKER

Subroutine CHCKER evaluates the first partial derivatives for all the functions at the starting point first by calling the user-supplied subroutine GRAD1 and then by calling DIFF1. The results are printed out to aid the user in debugging the subroutines used to describe the NLP problem he wishes to solve.

The matrix of second partial derivatives of each function is also evaluated by the two methods; first by calling MATRIX and then calling

FORTAN IV G LEVEL 21

BODY

DATE = 80240

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```

0001      SUBROUTINE BODY                                80000000
C                                                    80000001
C      AUGUST 1971                                     80000002
C                                                    80000003
C BODY COORDINATES THE FLOW AMONG THE SUBROUTINES THAT ACTUALLY DO THE 80000004
C CALCULATIONS REQUIRED BY THE VARIOUS PARTS OF THE ALGORITHM. 80000005
C IMPLICIT REAL*8(A-H,O-Z)                             80000006
0002      REAL*4 RHOIN,RATIO,EPSI,THETA0,XEP1,XEP2       80000007
0003      COMMON/SHARE/X(45),DEL(45),A(45,45),N,M,MN,NP1,NM1 80000008
0004      COMMON /CPTNS/ NT1,NT2,NT3,NT4,NT5,NT6,NT7,NT8,NT9,NT10 80000009
0005      COMMON/VALUE/F,G,PQ,RSIGMA,RJ(90),RHO         80000010
0006      COMMON/CRST/DELX(45),DELX0(45),RHOIN,RATIO,EPSI,THETA0, 80000011
0007      IRSIG1,G1,X1(45),X2(45),X3(45),XR2(45),XR1(45),PRI, 80000012
      2PR2,P1,F1,RJ1(90),DOTT,PGRAD(45),DIAG(45), 80000013
      3 PRFV3,ADELX, NTCTR, NUMINI, NPHASE, NSATIS 80000014
0008      COMMON /CONPAR/ NF1,NF2,NF3                   80000015
0009      COMMON/EXOPT/NEXP1,NEXP2,NEXP3,NEXP4,NEXP5,XEP1,XEP2 80000016
0010      COMMON/SEN/PAR(45),OPAR(45),NPAR,ISENS         80000017
0011      COMMON/ABG/LY,LZ,PER(45)                       80000018
0012      COMMON/ABG1/FE1,FE2                             80000019
0013      COMMON/ABG2/DF1(45),DF2(45)                   80000020
0014      NF2=2                                             80000021
0015      NF3=2                                             80000022
0016      MN=0                                              80000023
0017      NUMINI=0                                         80000024
0018      ISENS = 0                                         80000025
C                                                    80000026
C OPTION OF GETTING INITIAL RHO                        80000027
0019      CALL RHOCOM                                       80000028
0020      CALL EVALU                                         80000029
0021      CALL XMOVE                                         80000030
0022      GO TO (30,20), NT3                               80000031
0023      20 CALL TIMEC                                       80000032
0024      CALL OUTPUT (1)                                    80000033
0025      GO TO 40                                           80000034
0026      30 CALL TCHECK                                     80000035
C IN FEASIBILITY PHASE 4 MEANS FEAS ACHIEVED           80000036
0027      40 GO TO (50,50,50,200), NSATIS                 80000037
0028      50 CALL CONVRG (N1)                                80000038
0029      GO TO (60,10,125), N1                             80000039
C MINIMUM ACHIEVED IF N1=1                             80000040
0030      60 GO TO (70,80), NT3                             80000041
0031      70 CALL TIMEC                                       80000042
0032      CALL OUTPUT (1)                                    80000043
C --- NUMBER OF MINIMA ACHIEVED INCREASED BY 1         80000044
0033      80 NUMINI=NUMINI+1                                80000045
0034      MN=0                                              80000046
0035      GO TO (190,90,90), NPHASE                       80000047
0036      90 CALL ESTIM                                       80000048
C FINAL MIGHT HAVE BEEN CALLED BY ESTIM-CONVERGED IF N2=1 80000049
0037      IF(NEXP3.NE.2) GO TO 27                          80000050
0038      CALL SENS                                          80000051
0039      27 GO TO (100,110,120), NT4                      80000052
C NT4=1 FINAL CONVERGENCE ON 0 ORDER ESTIMATES, NT4=2 CONVERGE ON FIRST 80000053
C ORDER ESTIMATES, NT4=3 CONVERGE ON SECOND ORDER ESTIMATES. 80000054
0040      100 CALL FINAL (NF1)                              80000055
0041      GO TO (130,140), NF1                             80000056
0042      110 GO TO (130,140), NF2                         80000057
0043      120 GO TO (130,140), NF3                         80000058
0044      125 NPHASE=5                                       80000059
0045      130 RETURN                                         80000060
0046      140 RHO=RHO/RATIO                                  80000061
C USING PREVIOUSLY COMPUTED VALUES FOR F, AND RHO IS RECOMPUTED WITH THE 80000062
C NEW VALUE OF RHO.                                     80000063
0047      CALL PEVALU                                       80000064
C A VECTOR IS LEFT IN DELX(1) BY ESTIM                 80000065
0048      IF (NUMINI-2) 10,150,150                         80000066
0049      150 GO TO (110,160,160), NT7                     80000067
0050      160 CALL GRAD (2)                                   80000068
0051      CALL OPT                                           80000069
0052      GO TO (180,170), NT3                             80000070
0053      170 WRITE (6,210)                                  80000071
0054      CALL OUTPUT (1)                                    80000072
0055      180 GO TO 50                                       80000073
0056      190 IF (G) 90,90,200                             80000074
0057      200 RETURN                                         80000075
C --- DUAL VALUE GREATER THAN 0 MEANS NO FEASIBLE POINT EXISTS 80000076
C                                                    80000077
0058      210 FORMAT (6X,30MOVED ON EXTRAPOLATION VECTOR ) 80000078
0059      END                                                80000079

```

Figure 9.--Subroutine BODY.

**12/12/03**

Figure 10.--Subroutine BOUND.

```

      S, ' ESTIMATING F* ',/,1X,59(' '),//,11X,'F = ',E15.7,' * PAR(',
      $12,') + ',E15.7,///)
0069      WRITE(6,630) S1,LZ,B1
0070      630 FORMAT(2X,'LINE UNDER ESTIMATING F* AT POINT 1 ',/,1X,36(' '),
      $//,11X,'F = ',E15.7,' * PAR(',12,') + ',E15.7,///)
0071      WRITE(6,640) S2,LZ,B2
0072      640 FORMAT(2X,'LINE UNDER ESTIMATING F* AT POINT 2 ',/,1X,36(' '),
      $//,11X,'F = ',E15.7,' * PAR(',12,') + ',E15.7,///)
0073      WRITE(6,650) A,LZ,B,LZ,C
0074      650 FORMAT(2X,'QUADRATIC ESTIMATION OF F* THROUGH POINTS 1 AND'
      $, ' 2 ',/,1X,50(' '), //,11X,'F = ',E15.7,' * PAR(',12,
      $) ** 2 + ',E15.7,' * PAR(',12,') + ',E15.7,///)
0075      WRITE(6,660) LZ
0076      660 FORMAT(2X,'F* BOUND EVALUATION AT TEN EQUIDISTANCE POINTS',
      $, ' BETWEEN POINTS 1 AND 2 ',/,1X,71(' '),//,10X,'PAR(',12,')',
      $10X,'LOWER BOUND',10X,'UPPER BOUND',10X,'QUAD ESTIMATE',/,
      $10X,71(' '),9X,13(' '),8X,13(' '),8X,15(' '),/)
0077      WRITE(6,670) (E(I),FL12(I),FU12(I),FQ(I),I=1,10)
0078      WRITE(6,670) E2,FE2,FE2,FE2
0079      670 FORMAT(10X,F7.3,7X,E15.7,6X,E15.7,6X,E15.7)
0080      RETURN
C OUTPUT WHEN NEXOP7=2
0081      920 WRITE(6,620) S12,LZ,B12
0082      WRITE(6,660) LZ
0083      WRITE(6,921) (E(I),FU12(I), I=1,10)
0084      WRITE(6,921) E2,FE2
0085      921 FORMAT(10X,F7.3,28X,E15.7)
0086      RETURN
C OUTPUT WHEN NEXOP7=3
0087      930 WRITE(6,630) S1, LZ,B1
0088      WRITE(6,640) S2,LZ,B2
0089      WRITE(6,660) LZ
0090      WRITE(6,931) (E(I),FL12(I),I=1,10)
0091      WRITE(6,931) E2,FE2
0092      931 FORMAT(10X,F7.3,7X,E15.7)
0093      RETURN
C OUTPUT WHEN NEXOP7=4
0094      940 WRITE(6,941) S12,LZ,B12
0095      941 FORMAT(2X,'LINE PASSING THROUGH POINTS 1 AND 2 AND UNDER'
      $, ' ESTIMATING F* ',/,1X,59(' '),//,11X,'F = ',E15.7,' * PAR(',
      $12,') + ',E15.7,///)
0096      WRITE(6,942) S1,LZ,B1
0097      942 FORMAT(2X,'LINE OVER ESTIMATING F* AT POINT 1 ',/,1X,36(' '),
      $//,11X,'F = ',E15.7,' * PAR(',12,') + ',E15.7,///)
0098      WRITE(6,943) S2,LZ,B2
0099      943 FORMAT(2X,'LINE OVER ESTIMATING F* AT POINT 2 ',/,1X,36(' '),
      $//,11X,'F = ',E15.7,' * PAR(',12,') + ',E15.7,///)
0100      WRITE(6,650) A,LZ,B,LZ,C
0101      WRITE(6,660) LZ
0102      WRITE(6,670) (E(I),FU12(I),FL12(I),FQ(I),I=1,10)
0103      WRITE(6,670) E2,FE2,FE2,FE2
0104      RETURN
C OUTPUT WHEN NEXOP7=5
0105      950 WRITE(6,942) S1,LZ,B1
0106      WRITE(6,943) S2,LZ,B2
0107      WRITE(6,660) LZ
0108      WRITE(6,921) (E(I),FL12(I), I=1,10)
0109      WRITE(6,921) E2,FE2
0110      RETURN
C OUTPUT WHEN NEXOP7=6
0111      960 WRITE(6,941) S12,LZ,B12
0112      WRITE(6,660) LZ
0113      WRITE(6,931) (E(I),FU12(I), I=1,10)
0114      WRITE(6,931) E2,FE2
0115      RETURN
0116      END

```

Figure 10.--continued

DIFF2. Both results are printed out. If it is found that MATRIX creates a nonzero element below the main diagonal of A, then a switch is set to cause the statement STOP to be executed in MAIN. Subroutine CHCKER appears as Figure 11.

#### 6.4 CØNVRG

After each iteration of the algorithm to locate the minimum of the unconstrained function (the W function) subroutine CØNVRG is called to determine if the current point is an acceptable approximation of a point giving the minimum value of the W function. The argument N2 of this routine is given a value of 1 if the point is close enough; otherwise, it is given a value of 2. CØNVRG appears as Figure 12.

#### 6.5 DIFF1

Subroutine DIFF1 computes the first derivatives by numerical differencing. The user-supplied subroutine RESTNT is called 2n times for each gradient evaluated by DIFF1. For the function f, DIFF1 computes the *i*th component of the gradient using the formula

$$(\nabla f(x^0))_i = \frac{f(x^0 + \theta e_i) - f(x^0 - \theta e_i)}{2\theta},$$

where  $e_i$  is a vector of zeroes with a 1 in the *i*th component and  $\theta$  is a small number whose value is assigned by the user. DIFF1 is shown as Figure 13.

#### 6.6 DIFF2

Subroutine DIFF2 computes the second derivatives by numerical differencing. The matrix of second partials is computed using central differencing. The differences are calculated using the gradients of the function. The user-supplied subroutine GRAD1 is called 2n times for each matrix of second partial derivatives evaluated by DIFF2. DIFF2 is Figure 14.

[illegible]

**Figure 11.--Subroutine CHCKER.**

FORTRAN IV G LEVEL 21

CONVRG

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09/21/37

```

0001      SUBROUTINE CONVRG (N1)                                CON00040
C                                                    CON00050
C      OCTOBER 1970                                          CON00060
C                                                    CON00070
C AFTER EACH ITERATION OF THE ALGORITHM TO LOCATE THE MINIMUM OF THE CON00080
C PENALTY FUNCTION, CONVRG DETERMINES IF THE CURRENT POINT IS CLOSE CON00090
C ENOUGH TO THE POINT GIVING THE MINIMUM VALUE OF THE P FUNCTION. CON00100
C      N1 SET EQUAL TO 1 IF MINIMUM HAS BEEN FOUND.          CON00110
C      N1 SET EQUAL TO 2 IF MINIMUM HAS NOT BEEN FOUND AND TIME IS NOT UP CON00120
C      N1 SET EQUAL TO 3 OTHERWISE                            CON00130
C      DOTT SET EQUAL TO (DEL P)/(INVERSE(DEL(DEL P)))(DEL P) IN OPT CON00140
C      IMPLICIT REAL*8(A-H,O-Z)                               CON00150
0002      REAL*4 RHOIN,RATIO,EPSI,THETA0,XEP1,XEP2             CON00160
0003      COMMON/SHARE/X(45),DEL(45),AI(45,45),N,M,MN,NP1,NM1 CON00170
0004      COMMON /OPTNS/ NT1,NT2,NT3,NT4,NT5,NT6,NT7,NT8,NT9,NT10 CON00180
0005      COMMON/VALUE/F,G,PO,RSIGMA,RJ(90),RMO               CON00190
0006      COMMON/CRST/DELX(45),DELX0(45),RHOIN,RATIO,EPSI,THETA0, CON00200
0007      IRSIG1,G1,X1(45),X2(45),X3(45),XR2(45),XRI(45),PR1, CON00210
      2PK2,PI,F1,RJ1(90),DOTT,PGRAD(45),DIAG(45),           CON00220
      3 PREV3,ADELX, NTCTR, NUMINI, NPHASE, NSATIS           CON00230
      COMMON/EXOPT/NEXO1,NEXOP2,NEXOP3,NEXOP4,NEXOP5,XEP1,XEP2 CON00240
      COMMON /TSW/ NSW                                         CON00250
0010      N1=2                                                  CON00260
0011      IF (MN.LE.1) Q1=PO                                     CON00270
0012      GO TO (10,20,30), NT9                                CON00280
0013      10 IF(DABS(DOTT).LT.EPSI) GO TO 70                    CON00290
0014      GO TO 40                                               CON00300
0015      20 IF(DABS(DOTT).LT.(P1-PO)/5.0) GO TO 70             CON00310
0016      GO TO 40                                               CON00320
0017      30 IF (ADELX.LT.EPSI) GO TO 70                        CON00330
0018      GO TO (50,60), NSW                                     CON00340
0019      50 IF (MN.LE.1) RETURN                                  CON00350
0020      IF (PO+XEP2 .LT. Q1) GO TO 75                          CON00360
0021      WRITE (6,80)                                           CON00370
0022      GO TO 70                                               CON00380
0023      60 CALL PUNCH                                          CON00390
0024      WRITE (6,90)                                           CON00400
0025      N1=3                                                  CON00410
C                                                    CON00420
C      FOUND THE MINIMUM TO THE SUBPROBLEM.                   CON00430
C      RETURN                                                  CON00440
0026      70 N1=1                                               CON00450
0027      75 Q1 = PO                                             CON00460
0028      RETURN                                                 CON00470
C                                                    CON00480
0030      80 FORMAT (100H APPARENTLY ROUNDOFF ERRORS PREVENT A MORE ACCURATE DECON00490
      ITERATION OF THE MINIMUM OF THIS SUBPROBLEM.)           CON00500
0031      90 FORMAT (48H0*** TIME IS UP, CALLING EXIT FROM CONVRG. ***) CON00510
0032      END                                                    CON00520

```

Figure 12.--Subroutine CONVRG.

FORTRAN IV G LEVEL 21	DIFF1	DATE = 80242	09/22/00
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0001		SUBROUTINE DIFF1 (IN)	DIF00040
	C		DIF00050
	C	FEBUARY 1971	DIF00060
	C		DIF00070
	C	DIFF1 COMPUTES THE FIRST DERIVATIVES BY NUMERICAL DIFFERENCING.	DIF00080
	C		DIF00090
	C	--USER CAN CALL FOR DIFFERENCING OF SELECTED FUNCTIONS	DIF00100
0002		IMPLICIT REAL*8(A-H,O-Z)	DIF00110
0003		REAL*4 XEPI,XEP2	DIF00120
0004		COMMON/SHARE/X(45),DEL(45),A(45,45),N,N,MN,NP1,NM1	DIF00130
0005		COMMON/EXOPT/NEXOP1,NEXOP2,NEXOP3,NEXOP4,NEXOP5,XEPI,XEP2	DIF00140
0006		COMMON/STIRX/XSTR(45),XSSS(45),DOLL(45)	DIF00150
0007		DO 10 J=1,N	DIF00160
0008	10	XSTR(J)=X(J)	DIF00170
0009		DO 30 J=1,N	DIF00180
0010		IF (J.EQ.1) GO TO 20	DIF00190
0011		JM1=J-1	DIF00200
0012		X(JM1)=XSTR(JM1)	DIF00210
0013	20	X(J)=XSTR(J)+XEPI	DIF00220
0014		CALL RESTNT (IN,ZZ2)	DIF00230
0015		X(J)=XSTR(J)-XEPI	DIF00240
0016		CALL RESTNT (IN,ZZ1)	DIF00250
0017	30	DEL(J)=(ZZ2-ZZ1)/(2.*XEPI)	DIF00260
0018		X(N)=XSTR(N)	DIF00270
0019		RETURN	DIF00280
0020		END	DIF00290

Figure 13.--Subroutine DIFF1.



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DIFF2

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0001		SUBROUTINE DIFF2 (IN)	DIF00040
	C		DIF00050
	C	OCTOBER 1970	DIF00060
	C		DIF00070
	C	DIFF2 COMPUTES THE SECOND DERIVATIVES BY NUMERICAL DIFFERENCING.	DIF00080
	C		DIF00090
0002		IMPLICIT REAL*8(A-H,D-Z)	DIF00100
0003		REAL*4 XEP1,XEP2	DIF00110
0004		COMMON/SHARE/X(45),DEL(45),A(45,45),N,M,MN,NP1,NM1	DIF00120
0005		COMMON/EXPOPT/NEXOP1,NEXOP2,NEXOP3,NEXOP4,NEXOP5,XEP1,XEP2	DIF00130
0006		COMMON/STIAX/XSTR(45),XSSS(45),DOLL(45)	DIF00140
0007		DO 10 J=1,N	DIF00150
0008	10	XSSS(J)=X(J)	DIF00160
0009		DO 50 J=1,N	DIF00170
0010		IF (J.EQ.1) GO TO 20	DIF00180
0011		JM1=J-1	DIF00190
0012		X(JM1)=XSSS(JM1)	DIF00200
0013	20	X(J)=XSSS(J)+XEP1	DIF00210
0014		CALL GRAD1 (IN)	DIF00220
0015		DO 30 I=1,N	DIF00230
0016	30	DOLL(I)=DEL(I)	DIF00240
0017		X(I)=XSSS(I)-XEP1	DIF00250
0018		CALL GRAD1 (IN)	DIF00260
0019		DO 40 I=J,N	DIF00270
0020	40	A(I,J,I)=(DOLL(I)-DEL(I))/(2.*XEP1)	DIF00280
0021	50	CONTINUE	DIF00290
0022		X(IN)=XSSS(IN)	DIF00300
0023		RETURN	DIF00310
0024		END	DIF00320

Figure 14.--Subroutine DIFF2.

## 6.7 ESTIM

After the minimum of the  $W$  function for a given value of  $r$  has been located, ESTIM is called. When the minimum of the  $W$  function for a given value of  $r$  is determined a subproblem is said to be solved. ESTIM performs the computations to estimate the Lagrange multipliers and make the first- and second-order estimates of the final solution of the problem. ESTIM is Figure 15.

## 6.8 EVALU

In the normal phase, EVALU calls the user-supplied routines to evaluate the objective function and the constraint functions at the current point  $x$ . As the constraint functions are being computed, the code performs the calculations to compute the penalty terms of the penalty function. In the feasibility phase this routine puts the negative sum of the violated constraints, which is the objective function of the entry problem, in location  $F$ . In location  $F$  the current value of the objective function is stored for use by the program. After  $F$  has been computed, the value of the penalty function is computed. Also subroutine EVALU computes a value for the dual objective function. This value is only correct at the solution to a subproblem and the value is stored in the location labeled  $G$ . EVALU appears as Figure 16.

## 6.9 FEAS

Subroutine FEAS determines whether the starting point is an interior feasible point or not. If the variables of a problem are supposed to be nonnegative and some of the components of the starting point vector are nonpositive, then those components are changed to small positive numbers. A modification of FEAS must be made if it is desired to change this number. If the starting point does not satisfy the non-trivial constraints (the constraints other than the nonnegativity constraints), then the program goes into the feasibility phase. In this phase the negative of the sum of all the violated inequality constraints

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ESTIM

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```

0001      SUBROUTINE ESTIM
C
C      OCTOBER 1970
C
C ESTIM PERFORMS THE COMPUTATIONS TO ESTIMATE THE LAGRANGE MULTIPLIERS
C AND MAKE THE FIRST- AND SECOND-ORDER ESTIMATES OF THE FINAL SOLUTION
C OF THE PROBLEM.
0002      IMPLICIT REAL*8(A-H,O-Z)
0003      REAL*4 RHOIN,RATIO,EPSI,THETA0
0004      COMMON/SHARE/X(45),DEL(45),A(45,45),N,M,MN,NP1,NM1
0005      COMMON /EQAL/ H, H1, H2
0006      COMMON /OPTNS/ NT1,NT2,NT3,NT4,NT5,NT6,NT7,NT8,NT9,NT10
0007      COMMON/VALUE/F,G,PO,RSIGMA,RJ(90),RHO
0008      COMMON/CRST/DELX(45),DELY(45),RHOIN,RATIO,EPSI,THETA0,
1RSIG1,G1,X1(45),X2(45),X3(45),XR2(45),XR1(45),PR1,
2PR2,P1,F1,RJ1(90),DOTT,PGRAD(45),DIAG(45),
3 PREV3,ADDELX, NTCTR, NUMINI, NPHASE, NSATIS
0009      COMMON /COMPAR/ NFI,NF2,NF3
0010      CALL STORE
0011      Z10=RATIO**2
0012      Z9=RATIO
0013      Z1=1.0/Z9+1.0/Z10
0014      Z2=Z1+1./Z9**3
0015      Z3=1./Z9**3
0016      Z4=Z10+Z9
0017      Z5=Z3**3
0018      Z6=1.0/((Z10-1.0)*(Z9-1.0))
0019      Z7=1./Z9
0020      Z8=1./(Z9-1.)
0021      RQ=1.0/RHO
0022      IF (NUMINI-2) 150,80,10
0023      10  WRITE (6,330)
0024      PO=(PR2-Z4*PR1+Z5*P1)*Z6
0025      G=(RATIO*G1-GR1)/(RATIO-1.)
0026      DO 20 I=1,N
0027      20  X(I)=(XR2(I)-Z4*XR1(I)+Z5*X1(I))*Z6
0028      NP=NPHASE
0029      NPHASE=4
0030      CALL EVALU
0031      NPHASE=NP
0032      CALL OUTPUT (2)
C CHECK TO SEE IF ESTIMATES HAVE CONVERGED
0033      GO TO (70,30,70), NPHASE
0034      30  DO 50 J=1,M
0035      IF (RJ(J)) 40,50,50
0036      40  IF (THETA0+RJ(J)) 70,50,50
0037      50  CONTINUE
0038      GO TO (70,70,60), NT4
0039      60  CALL FINAL (NF3)
0040      70  CONTINUE
0041      80  WRITE (6,340)
0042      PO=(Z9*P1-PR1)*Z8
0043      G=(RATIO*G1-GR1)/(RATIO-1.)
0044      DO 90 I=1,N
0045      90  X(I)=(Z9*X1(I)-XR1(I))*Z8
0046      NP=NPHASE
0047      NPHASE=4
0048      CALL EVALU
0049      NPHASE=NP
0050      CALL OUTPUT (2)
C CHECK TO SEE IF ESTIMATES HAVE CONVERGED
0051      GO TO (140,100,140), NPHASE
0052      100 DO 120 J=1,M
0053      IF (RJ(J)) 110,120,120
0054      110 IF (RJ(J)+THETA0) 140,120,120
0055      120 CONTINUE
0056      GO TO (140,130,140), NT4
0057      130 CALL FINAL (NF2)
0058      140 CONTINUE
0059      150 WRITE (6,350)
0060      IF (M) 180,180,160
0061      160 DO 170 J=1,M
0062      RJ(J)=RHO/RJ1(J)
0063      170 IF (NZ) 210,210,190
0064      190 DO 200 J=1,MZ
0065      MNJ=M+J
0066      200 RJ(MNJ)=2.*RJ1(MNJ)*RQ
0067      210 GO TO (220,240), NT2
0068      220 DO 230 I=1,N
0069      230 X(I)=RHO/X1(I)
0070      240 CALL OUTPUT (2)
0071      CALL REJECT
0072      IF (NUMINI-2) 280,300,250
0073      250 GO TO (280,310,260), NT7
C SECOND ORDER MOVE FOR NEXT MINIMUM
0074      260 DO 270 I=1,N
0075      270 DELX(I)=Z1*X1(I)-Z2*XR1(I)+Z3*XR2(I)

```

Figure 15.--Subroutine ESTIM.

0076	280	PR2=PR1	EST00910
0077		GR2=GR1	EST00920
0078		PR1=P1	EST00930
0079		GR1=G1	EST00940
0080		DO 290 I=1,N	EST00950
0081		XR2(I)=XR1(I)	EST00960
0082	290	XR1(I)=X1(I)	EST00970
0083		RETURN	EST00980
0084	300	GO TO (280,310,310), NT7	EST00990
0085	310	DO 320 I=1,N	EST01000
0086	320	DELX(I)=(X1(I)-XR1(I))*Z7	EST01010
0087		GO TO 280	EST01020
	C		EST01030
0088	330	FORMAT (/26H0 2ND ORDER ESTIMATES )	EST01040
0089	340	FORMAT (/26H0 1ST ORDER ESTIMATES )	EST01050
0090	350	FORMAT (/25H0 LAGRANGE MULTIPLIERS )	EST01060
0091		END	EST01070

Figure 15.--continued

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EVALU

DATE = 80242

09/28/17

```

0001      SUBROUTINE EVALU                                EVA00043
C                                                    EVA00050
C      OCTOBER 1970                                EVA00063
C                                                    EVA00079
C IN THE NORMAL PHASE EVALU CALLS THE USER-SUPPLIED ROUTINES TO EVALUATE EVA00080
C THE OBJECTIVE FUNCTION AND THE CONSTRAINT FUNCTIONS AT THE CURRENT EVA00099
C POINT. IN THE FEASIBILITY PHASE THIS ROUTINE PUTS THE NEGATIVE SUM OF EVA00100
C THE VIOLATED CONSTRAINTS IN LOCATION F.                                EVA00110
0002      IMPLICIT REAL*8(A-H,O-Z)                                EVA00120
0003      REAL*4 RHOIN,RATIO,EPSI,THETA0                                EVA00130
0004      COMMON/SHARE/X(45),DEL(45),A(45,45),N,M,MN,NP1,NM1                                EVA00140
0005      COMMON /EQUAL/ H, H1, M2                                EVA00150
0006      COMMON /OPTNS/ NT1,NT2,NT3,NT4,NT5,NT6,NT7,NT8,NT9,NT10                                EVA00160
0007      COMMON/VALUE/F,G,PO,RSIGMA,RJ(50),RHO                                EVA00170
0008      COMMON/CRST/DELX(45),DELXO(45),RHOIN,RATIO,EPSI,THETA0,                                EVA00180
      IRSIG1,G1,X1(45),X2(45),X3(45),J2(45),XRI(45),PRI,                                EVA00190
      ZPR2,P1,F1,RJ1(90),DOTT,PGRAD(45),DIAG(45),                                EVA00200
      3 PREV3,ADELX, NTCTR, NUMINI, NPHASE, NSATIS                                EVA00210
0009      H=0.0                                EVA00220
0010      RSIGMA=0.0                                EVA00230
0011      F=0.0                                EVA00240
0012      NSATIS=2                                EVA00250
0013      GO TO (10,100,190,200), NPHASE                                EVA00260
C      =1 FEASIBILITY                                EVA00270
C      =2 NORMAL                                EVA00280
C      =3 GUESS                                EVA00290
C      =4 ALL FUNCTIONS ARE TO BE EVALUATED                                EVA00300
C FEASIBILITY                                EVA00310
0014      10 GO TO (20,40), NT2                                EVA00320
C NON-NEGATIVES INCLUDED                                EVA00330
0015      20 DO 30 I=1,N                                EVA00340
0016      IF (X(I)) 260,260,30                                EVA00350
0017      30 RSIGMA=RSIGMA-RHO*DLOG(X(I))                                EVA00360
0018      40 IF (M.EQ.0) GO TO 90                                EVA00370
0019      DO 80 J=1,M                                EVA00380
0020      CALL RESTNT (J,RJ(J))                                EVA00390
0021      IF (RJ(J).LE.0.0) GO TO 50                                EVA00400
0022      IF (RJ(J).GT.0.0) GO TO 60                                EVA00410
C VIOLATION OF A PREVIOUSLY SATISFIED CONSTRAINT                                EVA00420
0023      GO TO 260                                EVA00430
0024      50 IF (RJ(J).GT.0.0) GO TO 70                                EVA00440
C ALL VIOLATED CONSTRAINTS ADDED INTO OBJECTIVE FUNCTION                                EVA00450
0025      F=F-RJ(J)                                EVA00460
0026      GO TO 80                                EVA00470
0027      60 RSIGMA=RSIGMA-RHO*DLOG(RJ(J))                                EVA00480
0028      GO TO 80                                EVA00490
C INDICATES SATISFACTION OF CONSTRAINT(1ORMORE)                                EVA00500
0029      70 NSATIS=1                                EVA00510
0030      RSIGMA=RSIGMA-RHO*DLOG(RJ(J))                                EVA00520
0031      80 CONTINUE                                EVA00530
0032      90 CONTINUE                                EVA00540
C EQUALITIES NOT COMPUTED IN FEAS. PHASE                                EVA00550
0033      PO=F+RSIGMA                                EVA00560
0034      G=F-RHO*DFLOAT(M)                                EVA00570
0035      IF (NT2.EQ.1) G=G-RHO*DFLOAT(N)                                EVA00580
0036      RETURN                                EVA00590
C REGULAR PHASE                                EVA00600
0037      100 GO TO (110,130), NT2                                EVA00610
C NON NEGATIVITIES INCLUDED                                EVA00620
0038      110 DO 120 I=1,N                                EVA00630
0039      IF (X(I)) 260,260,120                                EVA00640
0040      120 RSIGMA=RSIGMA-RHO*DLOG(X(I))                                EVA00650
0041      130 IF (M.EQ.0) GO TO 150                                EVA00660
0042      DO 140 J=1,M                                EVA00670
0043      CALL RESTNT (J,RJ(J))                                EVA00680
0044      IF (RJ(J).LE.0.0) GO TO 260                                EVA00690
0045      RSIGMA=RSIGMA-RHO*DLOG(RJ(J))                                EVA00700
0046      140 CONTINUE                                EVA00710
C EVALUATE AND ADD IN EQUALITY CONSTRAINTS                                EVA00720
0047      150 CONTINUE                                EVA00730
0048      CALL RESTNT (O,F)                                EVA00740
0049      IF (M2) 180,180,160                                EVA00750
0050      160 DO 170 I=1,M2                                EVA00760
0051      J=I+M                                EVA00770
0052      CALL RESTNT (J,RJ(J))                                EVA00780
C ADD INTO THIRD TERM OF P FUNCTION                                EVA00790
0053      H=H+(RJ(J))**2                                EVA00800
0054      170 CONTINUE                                EVA00810
0055      H=H/RHO                                EVA00820
0056      180 PO=RSIGMA+H                                EVA00830
0057      PO=F+PO                                EVA00840
0058      G=2.*H-RHO*DFLOAT(M)                                EVA00850
0059      G=G+F                                EVA00860
0060      IF (NT2.EQ.1) G=G-RHO*DFLOAT(N)                                EVA00870
C DUAL VALUE                                EVA00880
0061      RETURN                                EVA00890
C GUESS PHASE NOT CODED                                EVA00900

```

Figure 16.-- Subroutine EVALU.

0062	190	RETURN	EVA00910
	C---	STRAIGHT FUNCTION EVALUATION ( MAIN+FEAS ONLY)	EVA00920
0063	200	CONTINUE	EVA00930
0064		IF (M.EQ.0) GO TO 220	EVA00940
0065		DO 210 I=1,M	EVA00950
0066		CALL RESTNT (I,RJ(I))	EVA00960
0067	210	CONTINUE	EVA00970
0068	220	CALL RESTNT (0,F)	EVA00980
	C	EQUALITY CONSTRAINTS	EVA00990
0069		IF (MZ) 250,250,230	EVA01000
0070	230	DO 240 I=1,MZ	EVA01010
0071		KZ=M+I	EVA01020
0072	240	CALL RESTNT (KZ,RJ(KZ))	EVA01030
0073	250	RETURN	EVA01040
	C	CONSTRAINTS VIOLATED NOT SO BEFORE	EVA01050
0074	260	NSATIS=3	EVA01060
0075		PO=10.E35	EVA01070
0076		RETURN	EVA01080
0077		END	EVA01090

Figure 16.--continued

is labeled for use as the objective function in the entry problem. Then the routine calls BODY to minimize this auxiliary function subject to the set of satisfied constraints. In the feasibility phase if a violated constraint is fortuitously satisfied an immediate return to FEAS is made and a new entry problem is begun, including the newly satisfied constraint in the set of constraints active for the entry problem. Thus the entry problem can result in a series of NLP problems being partly solved.

FEAS also contains tests that indicate when a problem does not contain a feasible starting point. Such information is printed out, and control is returned to MAIN with indicators set so that MAIN begins to try the next NLP problem in a stack of problems. When the entry problem is not a convex programming problem the test indicates only that the program is in a region from which it will be unable to locate a feasible starting point. The user, by supplying another starting point, may cause the algorithm to generate another sequence of points that does lead to a feasible starting point. Figure 17 shows FEAS.

#### 6.10 FINAL

Subroutine FINAL contains the test used to determine if a point satisfies the final convergence criterion of the algorithm. If a point does satisfy the criterion chosen by use of option 5 then N2, the argument of this routine, is given a value of 1; otherwise it is given a value of 2. FINAL is called following the computation of the solution estimates, which are made after the solution of the subproblem. FINAL appears as Figure 18.

#### 6.11 GRAD

The gradient of the W function is given by the formula

$$\nabla W(x,r) = \nabla f(x) - \sum_{i=1}^m \frac{r}{g_i(x)} \nabla g_i(x) + \sum_{i=m+1}^{m+p} \frac{2h_i(x)}{r} \nabla h_i(x) .$$

FORTRAN IV G LEVEL 21

FEAS

DATE = 80242

09/23/33

```

0001      SUBROUTINE FEAS                                FEA00040
C                                                    FEA00050
C                                                    FEA00060
C                                                    FEA00070
C FEAS DETERMINES WHETHER THE STARTING POINT IS FEASIBLE. IF IT IS NOT, FEA00080
C FEAS LOOKS FOR A FEASIBLE ONE. IF NONE EXISTS, A MESSAGE IS PRINTED FEA00090
C AND CONTROL RETURNS TO MAIN. FEA00100
C IMPLICIT REAL*8(A-H,O-Z) FEA00110
0002      REAL*8 RHOIN,RATIO,EPSI,THETA0 FEA00120
0003      COMMON/SHARE/X(45),DEL(45),A(45,45),N,M,MN,NP1,NM1 FEA00130
0004      COMMON /OPTMS/ NT1,NT2,NT3,NT4,NT5,NT6,NT7,NT8,NT9,NT10 FEA00140
0005      COMMON/VALUE/F,G,PO,RSIGNA,RJ(90),RMO FEA00150
0006      COMMON/CRST/DELX(45),DELX0(45),RHOIN,RATIO,EPSI,THETA0, FEA00160
0007      IRSIG1,G1,X1(45),X2(45),X3(45),XR2(45),XR1(45),PRI, FEA00170
      2PR2,P1,F1,RJ1(90),DOTT,PGRAD(45),DIAS(45), FEA00180
      3 PREV3,ADLX, NYCTR, NUMINI, NPHASE, NSATIS FEA00190
0008      NPHASE=1 FEA00200
0009      GO TO (10,90), NT2 FEA00210
0010      10 NFIX=1 FEA00220
0011      DO 30 I=1,N FEA00230
0012      IF (X(I)) 20,20,30 FEA00240
0013      20 NFIX=2 FEA00250
0014      X(I)=1.E-05 FEA00260
0015      30 CONTINUE FEA00270
0016      GO TO (50,40), NFIX FEA00280
0017      40 NPHASE=4 FEA00290
0018      CALL EVALU FEA00300
C JUST GET ALL CONSTRAINTS EVALUATED FEA00310
0019      NPHASE=1 FEA00320
0020      WRITE (6,130) FEA00330
0021      CALL OUTPUT (2) FEA00340
0022      50 IF (M) 90,90,60 FEA00350
0023      60 DO 70 I=1,M FEA00360
0024      IF (RJ(I)) 100,100,70 FEA00370
0025      70 CONTINUE FEA00380
0026      80 CALL TIMEC FEA00390
0027      WRITE (6,140) FEA00400
0028      G=0.0 FEA00410
0029      CALL RESTMT (0,F) FEA00420
0030      CALL OUTPUT (2) FEA00430
0031      90 RETURN FEA00440
0032      100 CALL BODY FEA00450
0033      IF(NPHASE .EQ. 5) RETURN FEA00460
0034      DO 110 I=1,M FEA00470
0035      IF (RJ(I)) 120,120,110 FEA00480
0036      110 CONTINUE FEA00490
0037      GO TO 80 FEA00500
0038      120 WRITE (6,150) FEA00510
C TO INDICATE TO MAIN TO START ON NEXT PROBLEM. FEA00520
0039      NPHASE=5 FEA00530
0040      GO TO 90 FEA00540
C FEA00550
0041      130 FORMAT (1H0,2X,48HMADE VIOLATED NON-NEGATIVITIES SLIGHTLY POSITIVEFEA00560
      1) FEA00570
0042      140 FORMAT (51H0****THE FEASIBLE STARTING POINT TO BE USED IS ...) FEA00580
0043      150 FORMAT (3X,89HTHIS PROBLEM POSSESSES NO FEASIBLE STARTING POINT, WFEA00590
      1ILL LOOK FOR DATA FOR NEXT PROBLEM. ) FEA00600
0044      END FEA00610

```

Figure 17.--Subroutine FEAS.



FORTRAN IV G LEVEL 21

FINAL

DATE = 80242

09/28/56

0001		SUBROUTINE FINAL (N2)	FIN00040
	C		FIN00050
	C	OCTOBER 1970	FIN00060
	C		FIN00070
	C	FINAL CONTAINS THE TESTS USED TO DETERMINE WHETHER A POINT SATISFIES	FIN00080
	C	THE FINAL CONVERGENCE CRITERION CHOSEN TO DETERMINE IF THE NLP	FIN00090
	C	PROBLEM HAS BEEN SOLVED.	FIN00100
	C	N2 SET EQUAL TO 1 IF CONVERGENCE CRITERION IS SATISFIED.	FIN00110
	C	N2 SET EQUAL TO 2 OTHERWISE.	FIN00120
0002		IMPLICIT REAL*8(A-H,O-Z)	FIN00130
0003		REAL*8 RHOIN,RATIO,EPSI,THETA0	FIN00140
0004		COMMON/SHARE/X(45),DEL(45),A(45,45),N,M,MN,NP1,NM1	FIN00150
0005		COMMON/VALUE/F,G,P0,RSIGMA,RJ(90),RMD	FIN00160
0006		COMMON/CRST/DELX(45),DELK0(45),RHOIN,RATIO,EPSI,THETA0,	FIN00170
		IRSIG1,G1,X1(45),X2(45),X3(45),YR2(45),XR1(45),PR1,	FIN00180
		ZPR2,P1,F1,RJ1(90),DOTT,PGRAD(45),DIAG(45),	FIN00190
		3 PREV3,ADELX, NTCTR, NUMINI, NPHASE, NSATIS	FIN00200
0007		GO TO (10,20,30), NT5	FIN00210
0008	10	EPSIL=DABS(F/G-1.)	FIN00220
0009		IF (EPSIL-THETA0) 50,50,70	FIN00230
0010	20	IF(DABS(RSIGMA)-THETA0) 50,50,70	FIN00240
0011	30	IF (NUMINI-1) 50,40,40	FIN00250
0012	40	PEST=PR1-(PR1-P0)/(1.-1./SQRT(RATIO))	FIN00260
0013		EPSIL=DABS(PEST/G-1.)	FIN00270
0014		IF (EPSIL-THETA0) 50,70,70	FIN00280
0015	50	N2=1	FIN00290
0016		GO TO (80,60), NT6	FIN00300
0017	60	CALL PUNCH	FIN00310
0018		GO TO 80	FIN00320
0019	70	N2=2	FIN00330
0020	80	RETURN	FIN00340
0021		END	FIN00350

Figure 18.--Subroutine FINAL.

Subroutine GRAD calls the user-supplied subroutine GRAD1 to compute  $\nabla f(x)$ ,  $\nabla g_i(x)$ , and  $\nabla h_i(x)$  and performs the computations to evaluate  $\nabla W(x,r)$ . The negative of the gradient of the  $W$  function (i.e.,  $-\nabla W$ ) is left in the array DELXO when GRAD returns control to the calling routine. When the argument of GRAD has a value of 2 this is all that is done. However, when it has a value of 1, GRAD also does part of the computations needed to evaluate the matrix of second partial derivatives of  $W$  at  $x$ . Subroutine GRAD is shown in Figure 19.

## 6.12 INVERS

When Newton's method is used to minimize the  $W$  function for a given value of  $r$ , it maps the negative of the gradient of  $W$  with the inverse of the matrix of second partial derivatives of  $W$  evaluated at  $x$ . A positive definite matrix will always give a vector along which the value of  $W$  will initially decrease. That is, for iteration  $i$ , a move is made along the vector  $S^i$ , given by the formula

$$S^i = -[\nabla^2 W(x,r)]^{-1} \nabla W(x,r).$$

This is equivalent to solving the set of simultaneous linear equations

$$[\nabla^2 W(x,r)] S^i = -\nabla W(x,r)$$

for  $S^{i*}$ , where  $\nabla^2 W$  and  $\nabla W$  are respectively the matrix of second derivatives and gradient vector of  $W$  evaluated at  $(x,r)$ .

Subroutine INVERS solves this set of equations for  $S^i$  using an L-U decomposition method (the Crout procedure). If it is determined that the matrix of second partials is not positive definite, a different procedure is used to obtain a direction  $S^i$ . A complete discussion of this procedure is given on page 167 of Fiacco and McCormick [10]. When

---

\*In the program the vector  $S^i$  is called DELX.

FORTRAN IV G LEVEL 21

GRAD

DATE = 80242

12/07/35

```

0001      SUBROUTINE GRAD (IS)                                GRA00040
C                                                    GRA00050
C              OCTOBER 1970                                GRA00060
C                                                    GRA00070
C GRAD COMPUTES THE GRADIENT OF THE PENALTY FUNCTION AND THE OUTER
C PRODUCT FACTORS OF THE MATRIX OF SECOND PARTIALS OF P.  GRA00080
C IF (IS=1) ALLOC. MATRIX OF 2ND PARTIALS IF (IS=2) DONT  GRA00090
0002      IMPLICIT REAL*8(A-H,O-Z)                            GRA00100
0003      REAL*8 RHOIN,RATIO,EPST,THETA0                      GRA00110
0004      COMMON/SHARE/X(45),DEL(45),A(45,45),N,M,MN,NP1,NM1  GRA00120
0005      COMMON /EQUAL/ H, M1, M2                            GRA00130
0006      COMMON /OPTNS/ NT1,NT2,NT3,NT4,NT5,NT6,NT7,NT8,NT9,NT10 GRA00140
0007      COMMON/VALUE/F,G,P0,RSIGMA,RJ(90),RHO              GRA00150
0008      COMMON/CRST/DELX(45),DELX0(45),RHOIN,RATIO,EPST,THETA0. GRA00160
1RSIG1,G1,X1(45),X2(45),X3(45),XR2(45),XR1(45),PR1,        GRA00170
2PR2,P1,F1,RJ1(90),DOIT,PGRAD(45),DIAG(45),                GRA00180
3 PREV3,ADLX, NCTR, NUMINI, NPHASE, NSATIS                  GRA00190
0009      GO TO (10,30), IS                                    GRA00200
0010      10 DO 20 I=1,N                                       GRA00210
0011          DO 20 J=1,I                                       GRA00220
0012          A(I,J)=0.                                         GRA00230
0013          DO 40 I=1,N                                       GRA00240
0014          40 DELX0(I)=0.                                     GRA00250
C THIS SECTION WORKS CORRECTLY IN FEASIBILITY PHASE AS WELL AS NORMAL PHASE GRA00260
0015      GO TO (50,80), NT2                                    GRA00270
0016      50 DO 70 I=1,N                                       GRA00280
0017          DELX0(I)=-RHO/X(I)                                GRA00290
0018          GO TO (60,70), IS                                  GRA00300
0019      60 A(I,I)=-DELX0(I)/X(I)                             GRA00310
0020      70 CONTINUE                                           GRA00320
0021      80 CONTINUE                                           GRA00330
0022          IF (M.LE.0) GO TO 180                             GRA00340
0023          DO 170 K=1,M                                       GRA00350
0024          CALL GRAD1 (K)                                     GRA00360
0025          IF (RJ(K).GT.0.0) GO TO 110                       GRA00370
C ALL VIOLATED CONSTRAINT GRADS ADDED TO OBJ. FUNCTION  GRA00380
0026      DO 100 I=1,N                                         GRA00390
0027          IF (DEL(I)) 90,100,90                             GRA00400
0028          DELX0(I)=DELX0(I)-DEL(I)                          GRA00410
0029      100 CONTINUE                                           GRA00420
0030          GO TO 170                                           GRA00430
0031      110 TT=RHO/RJ(K)                                       GRA00440
0032          DO 160 I=1,N                                       GRA00450
0033          IF (DEL(I)) 120,160,120                           GRA00460
C IF DEL(I)=0 SKIP ALL THE FOLLOWING COMPUTATION INVOLVING * BY DEL(I) GRA00470
0034      120 T=TT*DEL(I)                                       GRA00480
0035          DELX0(I)=DELX0(I)-T                                GRA00490
0036          GO TO (130,160), IS                                GRA00500
0037      130 T=T/RJ(K)                                         GRA00510
0038          DO 150 JJ=1,I                                       GRA00520
0039          IF (DEL(JJ)) 140,150,140                           GRA00530
0040          A(I,JJ)=A(I,JJ)+T*DEL(JJ)                         GRA00540
0041          150 CONTINUE                                       GRA00550
0042          160 CONTINUE                                       GRA00560
0043          170 CONTINUE                                       GRA00570
C EQUALITY CHANGES FOR GRAD                                GRA00580
0044      180 IF (MZ.LE.0) GO TO 250                            GRA00590
0045          GO TO (250,190,250), NPHASE                       GRA00600
0046      190 RQ=2./RHO                                         GRA00610
0047          DO 240 J=1,MZ                                       GRA00620
0048          K=M+J                                              GRA00630
0049          CALL GRAD1 (K)                                       GRA00640
0050          TT=RQ*RJ(K)                                         GRA00650
0051          DO 230 I=1,N                                       GRA00660
0052          IF (DEL(I).EQ.0.0) GO TO 230                       GRA00670
0053          DELX0(I)=DELX0(I)+DEL(I)*TT                       GRA00680
0054          GO TO (200,230), IS                                GRA00690
0055      200 T=RQ*DEL(I)                                       GRA00700
0056          DO 220 JJ=1,I                                       GRA00710
0057          IF (DEL(JJ)) 210,220,210                           GRA00720
0058          A(I,JJ)=A(I,JJ)+T*DEL(JJ)                         GRA00730
0059      220 CONTINUE                                           GRA00740
0060      230 CONTINUE                                           GRA00750
0061      240 CONTINUE                                           GRA00760
0062          GO TO (260,280), IS                                  GRA00770
0063          DO 270 I=1,N                                       GRA00780
0064          270 DIAG(I)=A(I,I)                                GRA00790
0065          GO TO (290,330,290), NPHASE                       GRA00800
C LEAVES NEGATIVE GRADIENT IN DELP                        GRA00810
0066      290 DO 300 I=1,N                                       GRA00820
0067          DELX0(I)=-DELX0(I)                                GRA00830
0068          310 ADLX=0.                                         GRA00840
0069          DO 320 I=1,N                                       GRA00850
0070          ADLX=ADLX+DELX0(I)**2                             GRA00860
0071          ADLX=DSQRT(ADLX)                                   GRA00870
0072          RETURN                                             GRA00880
0073      330 CALL GRAD1 (0)                                       GRA00890
0074          DO 340 I=1,N                                       GRA00900
0075          340 DELX0(I)=-DELX0(I)-DEL(I)                     GRA00910
C LEAVES THE NEG. GRAD OF P IN DELX0                      GRA00920
0076      GO TO 310                                             GRA00930
0077      END                                                    GRA00940

```

Figure 19.--Subroutine GRAD.

this occurs the program prints out "ORTHOGONAL MOVE." This also warns the user that the problem is probably not a convex program. Subroutine INVERS appears as Figure 20.

### 6.13 LMULT

Subroutine LMULT, called by subroutine SENS, calculates the Lagrange multiplier sensitivities according to steps 5 and 6 of Algorithm 2.1 (Section 2). LMULT is given in Figure 21.

### 6.14 MAIN

MAIN is the program that initiates the SENSUMT algorithm to solve a nonlinear programming (NLP) problem; it is not a subroutine. The input of parameters and options, as well as a starting point, is done in MAIN, and the call to READIN (a user-supplied subroutine) is made to allow the user to read in data needed to evaluate his objective function and constraint functions. Starting point data or blank cards should always be supplied by the user because starting point data cards are always read in MAIN. Subroutine FEAS is called to obtain a feasible starting point by making use of the SENSUMT algorithm to solve the entry problem if the user-supplied point is not feasible. The actual solution of the NLP problem using the SENSUMT algorithm is supervised by subroutine BODY. The calls to SENS (sensitivity subroutine) and BOUND (bound subroutine) are also done in MAIN. When calculating bounds, after the solution and sensitivity analysis of the unperturbed problem, MAIN reiterates SENSUMT to solve the subject problem with the perturbed data. Perturbations and readjustments in the problem data are done in PERT (perturbation subroutine), on call by MAIN. Subroutine MAIN is shown as Figure 22.

### 6.15 OPT

The purpose of subroutine OPT is to obtain a  $\bar{\theta} > 0$  such that  $w(\bar{x} + \bar{\theta}\bar{S})$  is a minimum with respect to  $\theta$  along the vector  $\bar{x} + \bar{\theta}\bar{S}$ .

```

FORTRAN IV G LEVEL 21          INVERS          DATE = 80242          12/20/20

0001          SUBROUTINE INVERS (NSME)                                INV00040
C                                                                    INV00050
C          OCTOBER 1970                                              INV00060
C                                                                    INV00070
C INVERS SOLVES THE SET OF EQUATIONS FOR THE MOVE-VECTOR USING THE INV00080
C CRUUT PROCEDURE. IF THE MATRIX IS NOT POSITIVE DEFINITE, A DIFFERENT INV00090
C METHOD IS USED.                                                  INV00100
C                                                                    INV00110
C****PERFORMING A L-U DECOMPOSITION OF THE MATRIX A, TAKING ADVANAGE OF INV00120
C****THE SYMMETRY OF THE A MATRIX.                                INV00130
C****IF A NON-POSITIVE PIVOT CANDIDATE IS GENERATED, THEN MCCORMICK,S INV00140
C****PROCEDURE IS USED(SEE PP. 167-168 IN FIACCO AND MCCORMICK). INV00150
C                                                                    INV00160
C****IF NSME =1 WORKING WITH A NEW A MATRIX. IF NSME= 2 USING PREVIOUS INV00170
C****A MATRIX, BUT HAVE A NEW RIGHT-HAND-SIDE.                    INV00180
C                                                                    INV00190
C****NINV IS THE NUMBER OF NON-POSITIVE PIVOT CANDIDATES GENERATED. INV00200
C      IMPLICIT REAL*8(A-H,O-Z)                                     INV00210
0002      REAL*4 RHOIN,RATIO,EPST,THETA0,XEPI,XEP2                    INV00220
0003      COMMON/SHARE/X(45),DEL(45),AI(45,45),N,M,MN,NP1,NM1      INV00230
0004      COMMON /OPTNS/ NT1,NT2,NT3,NT4,NT5,NT6,NT7,NT8,NT9,NT10  INV00240
0005      COMMON/CRST/DELX(45),DELX0(45),RHOIN,RATIO,EPST,THETA0,  INV00250
0006      IRSIG1,G1,X1(45),X2(45),X3(45),XR2(45),XR1(45),PR1,     INV00260
      ZPR2,P1,F1,RJ1(90),DOTT,PGRAD(45),DIAG(45),                INV00270
      3 PREV3,ADLX, NYCTR, NUMINI, NPHASE, NSATIS                INV00280
      COMMON/EXPOPT/NEXP01,NEXP02,NEXP03,NEXP04,NEXP05,XEPI,XEP2 INV00290
      DIMENSION B(45)                                           INV00300
0007      EQUIVALENCE (B,DELX)                                     INV00310
0008      GO TO(20, 170), NSME                                     INV00320
0009      20  NINV=0                                               INV00330
0010          IF (A(1,1)) 40,30,50                                INV00340
0011      30  NINV=1                                               INV00350
0012          GO TO 70                                             INV00360
0013      40  NINV=1                                               INV00370
0014          A(1,1)=1./A(1,1)                                     INV00380
0015          DO 60 I=2,N                                          INV00390
0016      50  A(1,I)=A(1,I)*A(1,1)                                INV00400
0017          DO 160 J=2,N                                          INV00410
0018      60  J=J-1                                                INV00420
0019          T=0.                                                 INV00430
0020          DO 90 I=1,JM1                                         INV00440
0021      70  IF (A(I,J)) 80,90,80                                INV00450
0022          T=T+A(I,I)*A(I,J)                                    INV00460
0023      80  CONTINUE                                             INV00470
0024          A(I,J)=A(I,J)-T                                       INV00480
0025          IF (A(J,J)) 110,100,120                              INV00490
0026      90  NINV=NINV+1                                           INV00500
0027          GO TO 170                                             INV00510
0028      100 NINV=NINV+1                                           INV00520
0029          A(J,J)=1./A(J,J)                                       INV00530
0030          IF (J.EQ.N) GO TO 170                                INV00540
0031      110 J=J+1                                                 INV00550
0032          DO 150 L=J+1,N                                         INV00560
0033      120  T=0.                                                 INV00570
0034          DO 140 I=1,JM1                                         INV00580
0035      130  IF (A(I,J)) 130,140,130                              INV00590
0036          T=T+A(I,I)*A(I,J)                                    INV00600
0037      140  CONTINUE                                             INV00610
0038          A(I,J)=A(I,J)-T                                       INV00620
0039          A(J,I)=A(J,I)*A(I,J)                                INV00630
0040          CONTINUE                                             INV00640
0041      150  CONTINUE                                             INV00650
0042      160  CONTINUE                                             INV00660
0043      170  CONTINUE                                             INV00670
0044          IF (NINV) 180,180,290                                INV00680
0045      180  B(1)=B(1)*A(1,1)                                     INV00690
0046          DO 210 J=2,N                                          INV00700
0047      190  T=0.                                                 INV00710
0048          J=J-1                                                 INV00720
0049          DO 200 I=1,JM1                                         INV00730
0050      200  IF (A(I,J)) 190,200,190                              INV00740
0051          T=T+A(I,I)*B(I)                                       INV00750
0052      210  CONTINUE                                             INV00760
0053          B(J)=(B(J)-T)*A(J,J)                                  INV00770
0054          CONTINUE                                             INV00780
0055          DO 240 I=1,NM1                                         INV00790
0056      220  MMK=M-I                                              INV00800
0057          DO 230 J=1,I                                          INV00810
0058      230  L=NP1-J                                             INV00820
0059          IF (A(MMK,L)) 220,230,220                             INV00830
0060      240  B(MMK)=B(MMK)-A(MMK,L)*B(L)                         INV00840
0061          CONTINUE                                             INV00850
0062          CONTINUE                                             INV00860
0063          GO TO (280,260), NT3                                    INV00870
0064      250  WRITE (6,430)                                       INV00880
0065      260  WRITE (6,420) (DELX0(I),I=1,N)                     INV00890
0066      270  WRITE (6,440)                                       INV00900
0067          WRITE (6,420) (DELX(I),I=1,N)
0068

```

Figure 20.--Subroutine INVERS.

0069	280	RETURN	INV00910
	C---	COMPUTE ORTHOGONAL MOVE	INV00920
0070	290	CONTINUE	INV00930
0071		DO 350 I=1,N	INV00940
0072		I=N-I+1	INV00950
0073		IF (A(I,1)) 310,300,320	INV00960
0074	300	B(I)=0.0	INV00970
0075		GO TO 350	INV00980
0076	310	B(I)=1.0	INV00990
0077		GO TO 330	INV01000
0078	320	B(I)=0.0	INV01010
0079	330	IPI=1+1	INV01020
0080		IF (IPI.GT.N) GO TO 350	INV01030
0081		DO 340 J=IPI,N	INV01040
0082	340	B(J)=B(I)-A(I,J)*B(I)	INV01050
0083	350	CONTINUE	INV01060
0084		GO TO 360	INV01070
	C--	CHECK MAYBE DO DIFF FOR P.S.D.	INV01080
0085	360	ZC2=0.0	INV01090
0086		DO 370 I=1,N	INV01100
0087	370	ZC2=ZC2+DELX0(I)*B(I)	INV01110
0088		IF (ZC2) 380,400,400	INV01120
0089	380	DO 390 I=1,N	INV01130
0090	390	B(I)=-B(I)	INV01140
0091	400	WRITE (6,450)	INV01150
	C MCC	ZANGWILL ONE MOD	INV01160
0092		IF (NEXP2.NE.2) GO TO 250	INV01170
0093		DO 410 K=1,N	INV01180
0094	410	B(K)=B(K)+DELX0(K)	INV01190
0095		GO TO 250	INV01200
	C		INV01210
0096	420	FORMAT (7E17.8)	INV01220
0097	430	FORMAT (1H0,6X,12HDEL P VECTOR)	INV01230
0098	440	FORMAT(1H0,6X,24HSECOND ORDER MOVE VECTOR)	INV01240
0099	450	FORMAT (1H0,6X,15HORTHOGONAL MOVE)	INV01250
0100		END	INV01260

Figure 20.--continued

FORTRAN IV G LEVEL 21

LMULT

DATE = 80242

12/00/23

```

0001      SUBROUTINE LMULT(IND,DELMU,DEM,DU)
C
C      1 MARCH 1976
C
C      SUBROUTINE LMULT IS USED TO ESTIMATE THE PARTIAL DERIVATIVES OF THE
C      LAGRANGE MULTIPLIERS. IT IS CALLED BY SENS WHEN NEXOP4 = 1. USING
C      THE DIFFERENTIABILITY OF X(I), IT TAKES ADVANTAGE OF THE RELATIONS
C      U(I) = RHO/G(I)**2 AND W(J) = 2*H(J)/RHO, DIFFERENTIATING THEM WITH
C      RESPECT TO THE PARAMETERS. THIS SUBROUTINE WAS CODED BY R.L. ARMACOST
0002      IMPLICIT REAL*8(A-H,D-Z)
0003      REAL*4 RHOIN,RATIO,EPST,THETA0,XEP1,XEP2
0004      COMMON/SHARE/X(45),DEL(45),A(45,45),N,M,MN,NP1,NM1
0005      COMMON/VALUE/F,G,PQ,RSGMA,RJ(90),RHO
0006      COMMON/EQUAL/H, H1, M2
0007      COMMON/CRST/DELX(45),DELX0(45),RHOIN,RATIO,EPST,THETA0,
      INTCR,NUMIN1,X1(45),X2(45),X3(45),XR2(45),XR1(45),PR1,
      2PR2,P1,F1,RJ1(90),DOTT,PGRAD(45),DIAG(45),
      3PREV3,ADDELX,RSIG1,G1,NPHASE,NSATIS
0008      DIMENSION DELMU(90),DU(90),KTEST(45)
0009      MMZ = M + M2
0010      GO TO (1,2,3,3), IND
C IND = 0
0011      DO 100 I=1,MMZ
0012      100  DELMU(I) = 0.
0013      RETURN
C IND = 1
0014      1  DO 50 I=1,MMZ
0015      50  DELMU(I) = RJ(I)
0016      RETURN
C IND = 2
0017      2  DO 60 I=1,MMZ
0018      60  DELMU(I) = (DELMU(I) - RJ(I))/DEM
0019      RETURN
0020      3  DO 70 I=1,MMZ
0021      CALL GRAD1(I)
0022      CALL RESTNT(I,VAL)
0023      SUM = 0.
0024      DO 71 JJ=1,N
0025      71  SUM = SUM + DEL(JJ)*DELX(JJ)
0026      IF(IND.EQ.4) GO TO 80
0027      DU(I) = -(SUM + DELMU(I))*RHO/(VAL**2)
0028      GO TO 70
0029      80  DU(I) = 2.*(SUM + DELMU(I))/RHO
0030      70  CONTINUE
0031      RETURN
0032      END

```

Figure 21.--Subroutine LMULT.

FORTMAN IV G LEVEL 21

MAIN

DATE = 80242

12/12/30

```

C
C      AUGUST 1971
C
C MAIN IS THE PROGRAM THAT INITIATES THE SUMT ALGORITHM. THE INPUT OF
C PARAMETERS, OPTIONS, AND STARTING POINT IS DONE IN MAIN. AFTER THE
C SOLUTION OF ONE NLP PROBLEM MAIN LOOKS FOR DATA FOR ANOTHER NLP PROB.
C MAIN HAS BEEN MODIFIED BY AHMACOST(1976) TO INCLUDE SENSITIVITY
C ROUTINES AND BY GHAEMI(1979) TO INCLUDE BOUND CALCULATIONS
0001      IMPLICIT REAL*8(A-H,O-Z)
0002      REAL*4 RHOIN,RATIO,EPSI,THETA0,XEP1,XEP2,TMMAX
0003      COMMON/SHARE/X(45),DEL(45),A(45,45),N,M,MN,NP1,NM1
0004      COMMON /EQUAL/ H, M1, M2
0005      COMMON /OPTNS/ NT1,NT2,NT3,NT4,N,5,NT6,NT7,NT8,NT9,NT10
0006      COMMON/VALUE/F,G,P0,RSIGMA,RJ(90),RHO
0007      COMMON/CRST/DELX(45),DELX0(45),RHOIN,RATIO,EPSI,THETA0,
      1RSIG1,G1,X1(45),X2(45),X3(45),XP2(45),XRI(45),PRI,
      2PR2,P1,F1,RJ1(90),DOTT,PGRAD(45),DIAG(45),
      3PREV3,DELX, NTCTR, NUMINI, NPHASE, NSATIS
0008      COMMON/SEN/PAR(45),DPAR(45),NPAR,ISENS
0009      COMMON/EXOPT/NEXOP1,NEXOP2,NEXOP3,NEXOP4,NEXOP5,XEP1,XEP2
0010      COMMON/ABG/LY,LZ,PER(45)
0011      COMMON/ABG1/FE1,FE2
0012      COMMON/ABG2/DF1(45),DF2(45)
0013      COMMON/OP6/NEXOP6
0014      COMMON/OP7/NEXOP7
0015      COMMON/ABG4/XY(45)
0016      COMMON/ABG6/XX(45)
0017      DIMENSION XZ(45)
C      PARAMETER CARD
0018      10 READ(5,50,END=40) EPSI,RHOIN,THEIA0,RATIO,TMMAX,M,N,MZ
0019      LY=0.
0020      LZ=0.
0021      RRR=RHOIN
0022      20 CONTINUE
0023      IF(LY.EQ. 0.) GO TO 15
0024      IF(NEXOP3.EQ. 0.) GO TO 40
0025      15 CALL SET (TMMAX)
C      INITIAL X VECTOR CARD FORMAT
0026      IF(LY.GT. 0.) GO TO 200
0027      READ (5,60) (X(I),I=1,N)
0028      DO 4000 I=1,N
0029      XZ(I)=X(I)
0030      4000 CONTINUE
0031      200 RHOIN=RRR
0032      NTCTR=0
0033      NP1=N+1
0034      NM1=N-1
C SUBROUTINE READIN IS UNDER PROGRAMMER CONTROL
0035      IF(LY.GT. 0.) GO TO 210
0036      CALL READIN
C OPTION CARD FOLLOWS PROGRAMMERS DATA
0037      READ (5,80) NT1,NT2,NT3,NT4,NT5,NT6,NT7,NT8,NT9,NT10
0038      GO TO 215
0039      210 CALL READIN
0040      CALL PERT
0041      IF(LZ.GT. NPAR) GO TO 40
0042      215 WRITE (6,110)
0043      WRITE (6,120) N,M,MZ,TMMAX,RHOIN,RATIO,EPSI,THETA0
0044      WRITE (6,130)
0045      WRITE (6,80) NT1,NT2,NT3,NT4,NT5,NT6,NT7,NT8,NT9,NT10
C--READ TOLERANCES
0046      IF(LY.GT. 0.) GO TO 220
0047      READ (5,60) XEP1, XEP2
0048      220 WRITE (6,90)
0049      WRITE (6,70) XEP1, XEP2
C--READ SECOND OPTION CARD
0050      IF(LY.GT. 0.) GO TO 230
0051      READ (5,80) NEXOP1,NEXOP2,NEXOP3,NEXOP4,NEXOP5,NEXOP6, NEXOP7
0052      230 WRITE (6,100)
0053      WRITE (6,80) NEXOP1,NEXOP2,NEXOP3,NEXOP4,NEXOP5,NEXOP6,NEXOP7
0054      CALL TIMEC
0055      NPHASE=4
C --- JUST TO GET AN INITIAL PRINTOUT
0056      CALL EVALU
0057      P0=0.0
0058      G=0.0
0059      H=0.0
0060      RSIGMA=0.0
0061      CALL OUTPUT (2)
0062      CALL STORE
0063      IF (NEXOP1.GT.1) CALL CMCKER
0064      IF (NEXOP1.EQ.3) STOP 01072
0065      IF (NEXOP1.EQ.5) STOP 01104
0066      CALL FEAS
C NPHASE=5 INDICATES EITHER TIME RAN OUT OR NO FEASIBLE POINT WAS FOUND
0067      GO TO (30,30,30,30,10), NPHASE
0068      30 NPHASE=7

```

Figure 22.--MAIN



0069	NTCTR=0	MAI00930
0070	CALL BUDDY	MAI00940
0071	IF(NPHASE.EQ.5) GO TO 17	MAI00950
0072	IF(NEXUP3.EQ.1) GO TO 18	MAI00960
0073	IF(NEXUP3.EQ.3) GO TO 19	MAI00970
	C FOLLOWING STATEMENT IS INCLUDED FOR WATER POLLUTION PROJECT ONLY	MAI00980
	C IT HAS TO BE REMOVED LATER (INT8=5 IMPLIES POLLUTION PROJECT)	MAI00990
0074	IF(INT8.EQ.5) CALL PROF	MAI01000
0075	GO TO 17	MAI01010
0076	18 ISENS = 4	MAI01020
0077	CALL SENS	MAI01030
0078	IF(LY.GT.0.) CALL BOUND	MAI01040
0079	LY=LY+1	MAI01050
0080	GO TO 17	MAI01060
0081	19 ISENS = 2	MAI01070
0082	DO 21 I=1,12	MAI01080
0083	CALL SENS	MAI01090
0084	IF(LY.GT.0.) CALL BOUND	MAI01100
0085	LY=LY+1	MAI01110
0086	21 CONTINUE	MAI01120
0087	17 DO 300 I=1,N	MAI01130
0088	X(I)=XZ(I)	MAI01140
0089	300 CONTINUE	MAI01150
0090	GO TO 20	MAI01160
0091	40 STOP	MAI01170
	C	MAI01180
	C PARAMETER CARD	MAI01190
0092	50 FORMAT (5E12.0,3I4)	MAI01200
	C INITIAL X VECTOR CARD FORMAT	MAI01210
0093	60 FORMAT (6E12.6)	MAI01220
0094	70 FORMAT (6E20.7)	MAI01230
	C OPTION CARD FORMAT	MAI01240
	C OPTION CARD FORMAT	MAI01250
0095	80 FORMAT (10I7)	MAI01260
0096	90 FORMAT(13H0 TOLERANCES )	MAI01270
0097	100 FORMAT (26H0 SECOND SET OF OPTIONS )	MAI01280
0098	110 FORMAT (56H1 NONLINEAR PROGRAMMING ROUTINE-SUMT VERSION 4 03/01/73)	MAI01290
	1 )	MAI01300
0099	120 FORMAT (1H0,5X,2HN=13.6X,2HN=13.6X,3HNZ=13//8X,10HMAX. TIME=E14.7,MAI01310	
	14X,2HR=E14.7,4X,6HRTIO=E14.7,6X,8HEPSILON=E14.7,4X,6HMETAT=E14.7)MAI01320	
0100	130 FORMAT (18H0 OPTIONS SELECTED)	MAI01330
0101	END	MAI01340

Figure 22.--continued

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A USER'S MANUAL FOR SENSUMT: A PENALTY FUNCTION COMPUTER PROGRA--ETC(U)  
OCT 80 A V FIACCO, A GHAEMI DAAG29-79-C-0062  
SERIAL-T-434 AR0-16229.9-M NL

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The vector  $\bar{S}$ , along which the search for the minimum is made, is determined in subroutine XMOVE. The method used to locate the minimum along the vector is the Golden Section search method, which is based on the Fibonacci search method (see [10, p. 193]).

The Golden Section method is a one-dimensional search method to find the minimum of the function that does not require the computation of derivatives. Figure 23 shows  $\emptyset$ PT.

#### 6.16 $\emptyset$ UTPUT

Subroutine  $\emptyset$ UTPUT contains most of the "write" statements used to print out information on the results of solving an NLP problem. It is used to print out information after each iteration and also to print out the solution estimates and the estimates of the Lagrange multipliers.  $\emptyset$ UTPUT is given in Figure 24.

#### 6.17 PARDIF

The subroutine PARDIF is called by subroutine SENS to assign a differencing increment for use in the central differencing formulas indicated in Steps 2 and 4 of Algorithm 2.1. It assigns a fixed value to the differencing interval when the sensitivity analysis is conducted at the final subproblem, or assigns a value over a specified interval for sensitivity analysis at the final subproblem. When a trajectory sensitivity analysis is performed, PARDIF returns a differencing interval which is dependent on the particular subproblem involved, refining the value as the subproblem approaches the final one. Subroutine PARDIF is shown in Figure 25.

#### 6.18 PERT

Subroutine PERT, called by the program MAIN, performs the book-keeping regarding various parametric changes and adjustments required in the process of bound calculation. Figure 26 is subroutine PERT.

FORTRAN IV G LEVEL 21

OPT

DATE = 80243

10/19/39

```

0001      SUBROUTINE OPT                                OPT00040
C                                                    OPT00050
C      MARCH 1971                                    OPT00060
C                                                    OPT00070
C OPT LOOKS FOR A MINIMUM ALONG THE SEARCH VECTOR USING THE GOLDEN    OPT00080
C SECTION SEARCH METHOD.                                     OPT00090
0002      IMPLICIT REAL*8(A-H,O-Z)                    OPT00100
0003      REAL*4 RHOIN,RATIO,EPSI,THETAO              OPT00110
0004      COMMON/SHARE/X(45),DEL(45),A(45,45),N,M,MN,NP1,NM1    OPT00120
0005      COMMON/VALUE/F,G,P0,KSIGMA,RJ(90),RHO        OPT00130
0006      COMMON/CRST/DELX(45),DELXO(45),RHOIN,RATIO,EPSI,THETAO,    OPT00140
      IRSIG1,G1,X1(45),X2(45),X3(45),XR2(45),XR1(45),PR1,          OPT00150
      ZPR2,P1,F1,RJ1(90),DOTT,PGRAD(45),DIAG(45),                OPT00160
      3 PKEV3,ADELX, NTCTR, NUMINI, NPHASE, NSATIS                OPT00170
0007      KSW=1                                          OPT00180
0008      N405=1                                          OPT00190
0009      P31=PU                                          OPT00200
0010      ISW=1                                          OPT00210
0011      DOTT=0.                                         OPT00220
0012      DO 10 J=1,N                                    OPT00230
0013      DOTT=DOTT+DELX(J)*DELXO(J)                    OPT00240
0014      GO TO 40                                         OPT00250
0015      2J DO 30 I=1,N                                  OPT00260
0016      3J DELX(I)=-DELX(I)                            OPT00270
0017      4J CONTINUE                                     OPT00280
0018      N404=J                                          OPT00290
0019      MN=MN+1                                          OPT00300
C MN IS NOW NUMB. OF POINTS AFTER MIN ACHIEVED          OPT00310
0020      NTCTR=NTCTR+1                                    OPT00320
0021      DO 50 I=1,N                                    OPT00330
0022      5J X2(I)=X(I)                                    OPT00340
0023      PX1=PO                                           OPT00350
0024      N401=0                                           OPT00360
0025      6J N401=N401+1                                    OPT00370
0026      DO 70 I=1,N                                    OPT00380
0027      7J X(I)=X2(I)+DELX(I)                            OPT00390
0028      CALL EVALU                                       OPT00400
C 1 MEANS SATIS.OF CONSTRAINT NT.PPEV. 2MEANS NOCHANGE 3MEANS VIOLATION OPT00410
C IF POINT IS NOT FEASIBLE GIVE IT AN ARBITRARILY HIGH VALUE      OPT00420
0029      GO TO (540,90,80), NSATIS                      OPT00430
0030      8J PX2=10.E35                                    OPT00440
0031      PU=10.E35                                        OPT00450
0032      GO TO 100                                        OPT00460
0033      9J CONTINUE                                     OPT00470
0034      PX2=PO                                           OPT00480
0035      IF (PX1-PX2) 100,100,150                      OPT00490
0036      10J IF (N401-2) 130,110,110                    OPT00500
0037      DO 120 I=1,N                                    OPT00510
0038      12J X1(I)=X(I)                                    OPT00520
0039      P1=PX2                                           OPT00530
0040      GO TO 430                                        OPT00540
C ONLY ONE POINT SO FAR COMPUTED                          OPT00550
0041      13J DO 140 I=1,N                                OPT00560
0042      14J X3(I)=X2(I)                                  OPT00570
0043      PREV3=PX1                                         OPT00580
0044      GO TO 180                                         OPT00590
0045      15J DO 160 I=1,N                                OPT00600
0046      X3(I)=X2(I)                                       OPT00610
0047      X2(I)=X(I)                                       OPT00620
0048      16J DELX(I)=1.61803399*DELX(I)                  OPT00630
0049      PREV3=PX1                                         OPT00640
0050      PX1=PX2                                           OPT00650
0051      GO TO 60                                          OPT00660
C GOLDEN SECTION SEARCH METHOD.                            OPT00670
C B VECTOR GOES TO X1(I)                                  OPT00680
0052      17J PO=1.E36                                    OPT00690
0053      N404=N404+1                                    OPT00700
0054      DO 190 I=1,N                                    OPT00710
0055      19J X1(I)=X(I)                                    OPT00720
0056      P1=PO                                           OPT00730
0057      DO 200 I=1,N                                    OPT00740
0058      X1(I)=.38196601*(X1(I)-X3(I))+X3(I)            OPT00750
0059      20J X2(I)=X(I)                                    OPT00760
0060      CALL EVALU                                       OPT00770
0061      GO TO (540,270,210), NSATIS                      OPT00780
0062      21J IF (N404.LT.30) GO TO 170                  OPT00790
0063      CONTINUE                                         OPT00800
C THERE IS NO REFERENCE TO 211, THE ABOVE STATEMENT IS A DUMMY STATEMENT OPT00810
C--IT IS POSSIBLE NO FEASIBLE POINT EXIST. IF NOT TRY MOVING ON DELXO. OPT00820
C-- IF IT IS NOT POSSIBLE TO MOVE ON DELXO THEN WE MUST BE AT A    OPT00830
C-- SOLUTION OF NLP PROBLEM.                                  OPT00840
0064      IF (N404.GT.100) GO TO 240                      OPT00850
0065      22J DO 23J I=1,N                                  OPT00860
0066      IF (DABS(X3(I))/X1(I))-1.)GT.1.E-7) GO TO 170    OPT00870
0067      23J CONTINUE                                     OPT00880
0068      24J GO TO (250,260), N405                      OPT00890

```

Figure 23.--Subroutine OPT.

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```

0069      250  N405=2
          C---TRY TO MOVE ON GRADIENT.
          NTCTR=NTCTR-1
          M4=M4-1
          GO TO 20
0070      260  WRITE (6,5HJ)
          CALL TIMEC
          CALL JUTPUT (1)
          CALL REJECT
          STOP 22042
          C
0074      270  CONTINUE
          N404=0
          PX1=PXJ
          DO 280 I=1,N
0077      280  X(I)=0.38196601*(X1(I)-X2(I))+X2(I)
          CALL EVALU
          GO TO (540,290,220), NSATIS
0081      290  PX2=PXJ
          N401=1
          N401=N401+1
          IF (N401-25) 340,310,310
0084      310  K3=2
          IF (N401-40) 320,460,460
          DO 330 I=1,N
0087      320  IF (DABS(X2(I)/X1(I)-1.0).GE.1.E-7) GO TO 340
          CONTINUE
          GO TO 460
0090      340  IF (DABS(PX1/PX2-1.).LE.1.E-7) GO TO 460
          IF (PX1-PX2) 350,460,400
          C FROM LEFTORRIGHT X3(I)=(PREV3)X2(I)(PX1)X(I)PX2 X1(I)P1
          350  DO 360 I=1,N
          360  X1(I)=X(I)
          C THRU AWAY RIGHT PART
          P1=PX2
          DO 370 I=1,N
          C POINTXP1 BECOMES XP2
          370  X(I)=.38196601*(X1(I)-X3(I))+X3(I)
          C TEMPORARILY IN X STORAGE
          CALL EVALU
          GO TO (540,380,170), NSATIS
0094      380  CONTINUE
          PX2=PX1
          C SWITCH VECTORS TO PROPER POSITION
          PX1=PXJ
          DO 390 I=1,N
          XX=X2(I)
          X2(I)=X(I)
          X(I)=XX
          390  GO TO 300
          C LEFT SIDE TOSSED AWAY
          C--- CHANGES FOR NONUNIMODAL FN
          C--- GO TO THRU AWAY RIGHT IN THIS CASE INIT VAL LT FIB PT
          400  IF (PREV3-PX2) 350,350,410
          410  DO 420 I=1,N
          X3(I)=X2(I)
          X2(I)=X(I)
          PREV3=PX1
          PX1=PX2
          420  DO 440 I=1,N
          X(I)=0.38196601*(X1(I)-X2(I))+X2(I)
          CALL EVALU
          GO TO (540,450,170), NSATIS
          450  CONTINUE
          PX2=PXJ
          GO TO 300
          C THE INTERIOR POINTS NOW GIVE EQUAL VALUE FOR P. COMPUTE MIDPOINT.
          460  DO 470 I=1,N
          DELXO(I)=X(I)
          X(I)=(DELXO(I)+X2(I))*0.5
          470  CONTINUE
          CALL EVALU
          GO TO (480,490), KSW
          IF (DABS(P0/PX1-1.).GT.1.E-7) GO TO 520
          480  GO TO (500,510), ISW
          490  IF (P0.LT.P31) GO TO 510
          500  ISW=2
          C IF P-FUNCTION DIDN'T GO DOWN TRY NEG VECT.
          GO TO 20
          510  RETURN
          520  DO 530 I=1,N
          530  X(I)=DELXO(I)
          GO TO 350
          C ARE WE NOW IN FEASIBILITY PHASE
          540  DO 550 I=1,M
          IF (KJ(I)) 560,560,550
          550  CONTINUE
          NSATIS=4
          RETURN
          C--- PROBLEM HAS BECOME FEASIBLE
          C--- P - FUNCTION CHANGES IF A CONSTRAINT BECOMES FEASIBLE
          560  MN=0
          DO 570 I=1,M
          RJ(I)=XJ(I)
          RETURN
          C
          580  FORMAT ( 80H OPT CAN'T FIND A FEASIBLE POINT, THAT GIVES A LOWER
          1ALUL OF THE P-FUNCTION. )
          END

```

OPT00900  
 OPT00910  
 OPT00920  
 OPT00930  
 OPT00940  
 OPT00950  
 OPT00960  
 OPT00970  
 OPT00980  
 OPT00990  
 OPT01000  
 OPT01010  
 OPT01020  
 OPT01030  
 OPT01040  
 OPT01050  
 OPT01060  
 OPT01070  
 OPT01080  
 OPT01090  
 OPT01100  
 OPT01110  
 OPT01120  
 OPT01130  
 OPT01140  
 OPT01150  
 OPT01160  
 OPT01170  
 OPT01180  
 OPT01190  
 OPT01200  
 OPT01210  
 OPT01220  
 OPT01230  
 OPT01240  
 OPT01250  
 OPT01260  
 OPT01270  
 OPT01280  
 OPT01290  
 OPT01300  
 OPT01310  
 OPT01320  
 OPT01330  
 OPT01340  
 OPT01350  
 OPT01360  
 OPT01370  
 OPT01380  
 OPT01390  
 OPT01400  
 OPT01410  
 OPT01420  
 OPT01430  
 OPT01440  
 OPT01450  
 OPT01460  
 OPT01470  
 OPT01480  
 OPT01490  
 OPT01500  
 OPT01510  
 OPT01520  
 OPT01530  
 OPT01540  
 OPT01550  
 OPT01560  
 OPT01570  
 OPT01580  
 OPT01590  
 OPT01600  
 OPT01610  
 OPT01620  
 OPT01630  
 OPT01640  
 OPT01650  
 OPT01660  
 OPT01670  
 OPT01680  
 OPT01690  
 OPT01700  
 OPT01710  
 OPT01720  
 OPT01730  
 OPT01740  
 OPT01750  
 OPT01760  
 OPT01770  
 OPT01780  
 OPT01790  
 OPT01800  
 OPT01810  
 OPT01820  
 OPT01830  
 OPT01840  
 OPT01850  
 OPT01860  
 OPT01870  
 OPT01880

Figure  
 23.--  
 continued

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FORTRAN IV G LEVEL 21          OUTPUT          DATE = 80242          12/12/50

0001      SUBROUTINE OUTPUT (K)                                OUT00060
C                                          OUT00070
C          OCTOBER 1970                                         OUT00080
C                                          OUT00090
C OUTPUT PRINTS OUT INFORMATION ON THE RESULTS OF EACH ITERATION AND THE OUT00100
C SOLUTION ESTIMATES AND THE ESTIMATES OF THE LAGRANGE MULTIPLIERS OUT00110
C THIS ROUTINE WAS MODIFIED BY GHAMEI(1979) WHILE INCORPORATING OUT00120
C BOUND CALCULATION TECHNIQUE. OUT00130
0002      IMPLICIT REAL*8(A-H,O-Z) OUT00140
0003      REAL*4 RHOIN,RATIO,EPSI,THETA0 OUT00150
0004      COMMON/SHARE/X(45),DEL(45),A(45,45),N,M,MN,NP1,NM1 OUT00160
0005      COMMON /EQUAL/ M, M1, M2 OUT00170
0006      COMMON /OPTNS/ NT1,NT2,NT3,NT4,NT5,NT6,NT7,NT8,NT9,NT10 OUT00180
0007      COMMON/VALUE/F,G,PO,RSIGMA,RJ(90),RHO OUT00190
0008      COMMON/CRST/DELX(45),DELX0(45),RHOIN,RATIO,EPSI,THETA0, OUT00200
      1RSIG1,G1,X1(45),X2(45),X3(45),XR2(45),XR1(45),PRI, OUT00210
      2PR2,P1,F1,RJ1(90),DOTT,PGRAD(45),DIAG(45), OUT00220
      3PREV3,ADELX, NTCTR, NUMIN1, NPHASE, NSAVIS OUT00230
0009      COMMON/ABG/LY,LZ,PER(45) OUT00240
0010      COMMON/ABG1/FE1,FE2 OUT00250
0011      COMMON/ABG4/XV(45) OUT00260
0012      COMMON/ABG2/DF1(45),DF2(45) OUT00270
0013      COMMON/OP6/NEXOP6 OUT00280
0014      COMMON/ABG5/TX(45),TOELX(45) OUT00290
0015      COMMON/ABG6/XX(45) OUT00300
0016      NZ=M+M2 OUT00310
0017      GO TO (10,20), K OUT00320
0018      10 WRITE (6,60) OUT00330
0019      WRITE (6,70) NTCTR,DOTT,RHO,ADELX,NPHASE OUT00340
0020      20 WRITE (6,80) F,PO,G,RSIGMA,M OUT00350
0021      IF(LY.GT. 0.) GO TO 200 OUT00360
0022      DO 210 I=1,N OUT00370
0023      XV(I)=X(I) OUT00380
0024      210 CONTINUE OUT00390
0025      FE1=F OUT00400
0026      GO TO 220 OUT00410
0027      200 FE2=F OUT00420
0028      220 WRITE (6,90) (X(I),I=1,N) OUT00430
C CALCULATION OF GP VARIABLES OUT00440
0029      IF(NEXOP6.NE.1) GO TO 315 OUT00450
0030      INOEK=0. OUT00460
0031      CALL TRANS(INOEK) OUT00470
0032      WRITE(6,300)(X(I),I=1,N) OUT00480
0033      DO 410 I=1,N OUT00490
0034      410 X(I)=TX(I) OUT00500
0035      315 WRITE (6,110) OUT00510
0036      GO TO (30,40), MT2 OUT00520
0037      30 WRITE (6,120) OUT00530
0038      WRITE (6,100) (RJ(I),I=1,NZ) OUT00540
0039      GO TO 50 OUT00550
0040      40 WRITE (6,100) (RJ(I),I=1,NZ) OUT00560
0041      300 FORMAT(/,6X,'CORRESPONDING VALUE OF GP VARIABLE IS',/(6E20.7)) OUT00570
0042      50 RETURN OUT00580
C OUT00590
0043      60 FORMAT (50H0***** ) OUT00600
0044      70 FORMAT (10X,6HPDINT=14,6X,6H DOTT=E15.7,6X,4HRHO=E15.7,6X,10HMAGNI OUT00610
      1TUDE=E15.7,6X,6HPHASE=12) OUT00620
0045      80 FORMAT (8X,2HF=E15.7,5X,2HP=E15.7,5X,2MG=E15.7,5X,7HRSIGMA=E15.7,5 OUT00630
      1X,2HM=E15.7) OUT00640
0046      90 FORMAT (6X,25HTHE CURRENT VALUE OF X IS/(6E20.7)) OUT00650
0047      100 FORMAT (6E20.7) OUT00660
0048      110 FORMAT (6X,21HTHE CONSTRAINT VALUES) OUT00670
0049      120 FORMAT (28X,34HNOT INCLUDING THE NON-NEGATIVITIES) OUT00680
0050      END OUT00690

```

Figure 24.--Subroutine OUTPUT.

```

FORTRAN IV G LEVEL 21          PARDIF          DATE = 80242          12/09/35

0001          SUBROUTINE PARDIF                                PAR00340
C                                                              PAR00050
C              15 MARCH 1972                                PAR00060
C                                                              PAR00070
C THIS SUBROUTINE RETURNS A DIFFERENCING INTERVAL FOR DPAR(20) WHEN
C CALLED BY SUBROUTINE SENS. ISENS CONTROLS THE VALUE OF THE INTERVAL
C WITH A MINIMUM VALUE IN THE NEIGHBORHOOD OF 1.E-15 DEPENDING ON XE1.
C ISENS DEPENDS ON THE KIND OF SENSITIVITY ANALYSIS BEING CONDUCTED.
C THE SUBROUTINE WAS CODED BY R. L. ARMACOST.
C                                                              PAR00080
C                                                              PAR00090
C                                                              PAR00100
C                                                              PAR00110
C                                                              PAR00120
C                                                              PAR00130
C                                                              PAR00140
C                                                              PAR00150
0002          IMPLICIT REAL*8(A-H,O-Z)                        PAR00160
0003          REAL*4 XE1                                       PAR00170
0004          COMMON/EXOPT/NEXOP1,NEXOP2,NEXOP3,NEXOP4,NEXOP5,XE1,XE2
0005          COMMON/SEN/PAR(45),DPAR(45),NPAR,ISENS            PAR00180
0006          MULT = MIN0(ISENS,8)                             PAR00190
0007          XE1 = XE1 / (10.**MULT)                           PAR00200
0008          IF(XE1.LE.0.) XE1 = 1.E-10                       PAR00210
0009          DO 10 I=1,NPAR                                    PAR00220
0010          10 DPAR(I)=XE1                                    PAR00230
0011          RETURN                                           PAR00240
0012          END

```

Figure 25.--Subroutine PARDIF.

```

FORTRAN IV G LEVEL 21          PERT          DATE = 80242          12/13/33

C SUBROUTINE PERT IS CODED BY GHAEMI(1979) TO PERFORM THE PARAMETER
C ADJUSTMENTS REQUIRED AT VARIOUS STAGES OF BOUND CALCULATIONS.
C
0001          SUBROUTINE PERT                                PER00060
0002          IMPLICIT REAL*8(A-H,O-Z)                        PER00070
0003          COMMON/ABG/LY,LZ,PER(45)                       PER00080
0004          COMMON/SEN/PAR(45),DPAR(45),NPAR,ISENS          PER00090
0005          20 LZ=LZ+1                                       PER00100
0006          IF(LZ.GT. NPAR) GO TO 30                         PER00110
0007          IF(PER(LZ).NE. 0.) GO TO 10                      PER00120
0008          GO TO 20                                          PER00130
0009          10 PAR(LZ)=PAR(LZ)+PER(LZ)                       PER00140
0010          30 RETURN                                         PER00150
0011          END                                              PER00160

```

Figure 26.--Subroutine PERT.

## 6.19 PEVALU

Subroutine PEVALU is used to compute the value of the  $w$  function and the  $G$  function. It makes use of the values of the objective function and the constraint function, which must have been computed before PEVALU is called. The  $G$  function is interpreted as the value of the dual objective function when the  $W$  function has been minimized for a given value of  $r$  ( $RH0$ ). PEVALU is given as Figure 27.

```

FORTRAN IV G LEVEL 21          PEVALU          DATE = 80242          12/11/19

0001          SUBROUTINE PEVALU                                PEV00040
      C                                PEV00050
      C                                PEV00060
      C                                PEV00070
      C PEVALU COMPUTES THE VALUE OF THE PENALTY FUNCTION AND THE VALUE OF THE PEV00080
      C DUAL USING PREVIOUSLY COMPUTED VALUES FOR F, AND RJ.          PEV00090
0002          IMPLICIT REAL*8(A-H,O-Z)                        PEV00100
0003          REAL*4 RHOIN,RATIO,EPSI,THETA0                  PEV00110
0004          COMMON/SHARE/X(45),DEL(45),A(45,45),N,M,MN,NP1,NM1 PEV00120
0005          COMMON /EQAL/ H, M1, M2                         PEV00130
0006          COMMON /OPTNS/ NT1,NT2,NT3,NT4,NT5,NT6,NT7,NT8,NT9,NT10 PEV00140
0007          COMMON/VALUE/F,G,P0,RSIGMA,RJ(90),RHO           PEV00150
0008          COMMON/CRST/DELX(45),DELX0(45),RHOIN,RATIO,EPSI,THETA0, PEV00160
          1RSIG1,G1,X1(45),X2(45),X3(45),XR2(45),XR1(45),PR1, PEV00170
          2PR2,P1,F1,RJ1(90),DOIT,PGRAD(45),DIAG(45), PEV00180
          3 PREV3,ADLX, NTCTR, NUMINI, NPHASE, NSATIS PEV00190
0009          H=0.0                                           PEV00200
0010          RSIGMA=0.0                                       PEV00210
      C NONVEGS IF INCLUDED ARE ADDED TO P--ARE PDS. IN ALL PHASES PEV00220
0011          GO TO (10,30), NT2                               PEV00230
0012          DO 20 I=1,N                                       PEV00240
0013          20 RSIGMA=RSIGMA-RHO*DLOG(X(I))                 PEV00250
0014          30 GO TO (40,50,150), NPHASE                     PEV00260
      C OBJECTIVE FUNCTION - SIGMA VIOL. CONSTS.              PEV00270
0015          40 F=0.0                                          PEV00280
0016          50 IF (M) 100,100,60                             PEV00290
0017          60 DO 90 J=1,M                                     PEV00300
0018          IF (RJ(J)) 80,80,70                               PEV00310
0019          70 RSIGMA=RSIGMA-RHO*DLOG(RJ(J))                 PEV00320
0020          GO TO 90                                           PEV00330
0021          80 F=F-RJ(J)                                       PEV00340
0022          90 CONTINUE                                       PEV00350
      C EQUALITIES NOT ADDED IN FEAS. PHASE                    PEV00360
0023          100 CONTINUE                                      PEV00370
0024          IF (M2) 140,140,110                               PEV00380
0025          110 GO TO (140,120,150), NPHASE                  PEV00390
0026          120 DO 130 I=1,M2                                  PEV00400
0027          K=M+I                                              PEV00410
0028          130 H=H+RJ(K)**2                                   PEV00420
0029          H=H/RHO                                           PEV00430
0030          140 HS=H+RSIGMA                                    PEV00440
0031          PO=F+HS                                             PEV00450
0032          HMS=2.*H-RHO*DFLOAT(M)                            PEV00460
0033          G=F+HMS                                            PEV00470
0034          IF (NT2.EQ.1) G=G-RHO*DFLOAT(N)                   PEV00480
0035          150 RETURN                                         PEV00490
0036          END                                               PEV00500

```

Figure 27.--Subroutine PEVALU.

## 6.20 PRESEN

This subroutine calculates the gradient of the optimal value function, using the gradient of the Lagrangian when taken with respect to the parameters (Step 9 of Algorithm 2.1). In addition, PRESEN (when invoked) performs a preliminary screening of the problem parameters to which the optimal value function is practically insensitive. The criterion used for this purpose is that if the change in optimal value function is less than 0.1 percent of its current value as a result of the introduced increment of a given parameter, then that parameter is



eliminated from further consideration. This parameter screening is invoked when the fifth variable in the second option card takes a value of zero, as shown in Table 4. Figure 28 lists PRESEN.

#### 6.21 PUNCH

When a call to PUNCH is made, the current value of the array  $x$  is punched, along with some of the control cards that can be used to restart SENSUMT. Subroutine PUNCH is shown in Figure 29.

#### 6.22 REJECT

Subroutine REJECT puts values of the point at  $x$  and the associated values of the objective function, constraint functions, and  $W$  function back into their normal locations. These values were temporarily stored in other locations by subroutine STORE. STORE appears as Figure 30.

#### 6.23 RHOCOM

This subroutine is used to compute the initial value of  $r$  (RH0) for each NLP problem. As previously stated, the search for a feasible starting point can result in a sequence of NLP problems. The initial value of  $r$  is computed by the formula specified for the use of option 1 (NT1). RHOCOM is given as Figure 31.

#### 6.24 SECOND

SECOND is used to query about the computer's clock. This is made possible by an internal clock which is called by this subroutine. The elapsed time obtained and monitored by this subroutine is used by the subroutines TIMEC, TCHECK, and SET. Figure 32 is subroutine SECOND.

FORTRAN IV G LEVEL 21

PRESEN

DATE = 80242

12/13/50

```

0001      SUBROUTINE PRESEN(DU,KTEST)
C
C          1 MARCH 1976
C
C SUBROUTINE PRESEN IS USED TO CALCULATE THE GRADIENT OF THE OPTIMAL
C VALUE FUNCTION USING THE GRADIENT OF THE LAGRANGIAN TAKEN WITH
C RESPECT TO THE PARAMETERS. PRESEN IS CALLED BY SENS WHEN VARIABLE
C NEXOPS = 0 OR = 1. ADDITIONALLY, WHEN NEXOPS = 0, THE PARAMETERS WHICH
C AFFECT THE OPTIMAL VALUE FUNCTION BY LESS THAN 0.001 TIMES ITS
C CURRENT VALUE ARE ELIMINATED FROM FURTHER CONSIDERATION. THIS
C SUBROUTINE WAS CODED BY R. L. ARMACOST.
0002      IMPLICIT REAL*8(A-M,D-Z)
0003      COMMON/SHARE/X(45),DEL(45),A(45,45),N,M,MN,NP1,NM1
0004      COMMON/VALUE/F,G,PO,RSIGMA,RJ(90),RHO
0005      COMMON/EQUAL/ H, M1,MZ
0006      COMMON/SEN/PAR(45),DPAR(45),NPAR,ISENS
0007      COMMON/ABG/LY,LZ,PER(45)
0008      COMMON/ABG1/FE1,FE2
0009      COMMON/ABG2/DF1(45),DF2(45)
0010      DIMENSION GX(90),DU(90),KTEST(45),KLIST(45)
0011      MPHZ = M + MZ
0012      FTEST = 0.001 * DABS(F)
0013      DO 100 J=1,NPAR
0014      KTEST(J) = 0
0015      PAR(J) = PAR(J) + DPAR(J)
0016      CALL RESTNT(0,DF)
0017      IF(MPHZ.EQ.0) GO TO 20
0018      DO 10 I=1,MPHZ
0019      CALL RESTNT(1,DU(I))
0020      10 CONTINUE
0021      20 DEM = 2. * DPAR(J)
0022      PAR(J) = PAR(J) - DEM
0023      CALL RESTNT(0,XF)
0024      IF(MPHZ.EQ.0) GO TO 40
0025      DO 30 I=1,MPHZ
0026      CALL RESTNT(1,GX(I))
0027      30 CONTINUE
0028      40 DFEP5 = (DF - XF)/DEM
0029      IF(MPHZ.EQ.0) GO TO 60
0030      DO 50 I=1,MPHZ
0031      DU(I) = (DU(I) - GX(I))/DEM
0032      60 SUM = DFEP5
0033      IF(M.EQ.0) GO TO 80
0034      DO 70 I=1,M
0035      SUM = SUM - RHO/RJ(11)*DU(I)
0036      70 IF(MZ.EQ.0) GO TO 95
0037      TSUM = 0.
0038      DO 90 I=1,MZ
0039      IM = I+M
0040      TSUM = TSUM + RJ(IM)*DU(IM)
0041      SUM = SUM + TSUM * 2./RHO
0042      95 DEL(J) = SUM
0043      PAR(J) = PAR(J) + DPAR(J)
0044      DTEST = DABS(DEL(J))
0045      IF(DTEST.GE.FTEST) KTEST(J) = 1
0046      100 CONTINUE
0047      WRITE(6,600)
0048      600 FORMAT(22X,34HOPTIMAL VALUE FUNCTION SENSITIVITY    //)
0049      DO 200 I=1,NPAR,5
0050      II=MINO(I+4,NPAR)
0051      WRITE(6,601) ((JJ,DEL(JJ)), JJ=I,II)
0052      601 FORMAT( 51H DF/DA(,I2,2H)=,G14.7)
0053      200 CONTINUE
0054      IF(LY.GT. 0.) GO TO 500
0055      DO 400 J=1,NPAR
0056      DF1(J)=DEL(J)
0057      400 CONTINUE
0058      GO TO 700
0059      500 DO 610 J=1,NPAR
0060      DF2(J)=DEL(J)
0061      610 CONTINUE
0062      IF(LY.EQ. 0.) GO TO 700
0063      PAR(LZ)=PAR(LZ)-PER(LZ)
0064      RETURN
0065      700 JJ = 0
0066      DO 250 J=1,NPAR
0067      IF(KTEST(J).EQ.0) GO TO 250
0068      JJ = JJ + 1
0069      KLIST(JJ) = J
0070      250 CONTINUE
0071      IF(JJ.EQ.0) GO TO 300
0072      WRITE(6,602)
0073      602 FORMAT( /31H DETAILED SENSITIVITY RESULTS FOLLOW FOR PARAMETERS )
0074      WRITE(6,603) (KLIST(I), I=1,JJ)
0075      603 FORMAT(1H 40(I2,2H ,))
0076      WRITE(6,604)
0077      604 FORMAT(/)
0078      RETURN
0079      300 WRITE(6,605)
0080      605 FORMAT( 42H THERE ARE NO DETAILED SENSITIVITY RESULTS    //)
0081      RETURN
0082      END

```

Figure 28.--Subroutine PRESEN.

```

FURTKAN IV G LEVEL 21          PUNCH          DATE = 80242          12/12/21

0001      SUBROUTINE PUNCH                      PUN00040
C                      OCTOBER 1970              PUN00050
C                      C THIS SUBROUTINE PUNCHES THE STOPPING POINTS AND ASSOCIATED PARAMETERS
C                      C SO THAT ANOTHER RUN MAY BE MADE STARTING WHERE THE CUKRENT ONE
C                      C STOPPED                PUN00060
C                      C THIS ROUTINE IS CALLED IF NT6=2.
C                      C                      PUN00070
0002      IMPLICIT REAL*8(A-H,O-Z)              PUN00080
0003      REAL*4 RHOIN,RATIO,EPST,THETA0,XEP1,XEP2,T PUN00090
0004      COMMON/SHARE/X(45),DEL(45),A(45,45),N,M,MN,NP1,NM1 PUN00100
0005      COMMON /EQUAL/ H, H1, H2              PUN00110
0006      COMMON /OPTNS/ NT1,NT2,NT3,NT4,NT5,NT6,NT7,NT8,NT9,NT10 PUN00120
0007      COMMON/VALUE/F,G,PO,RSIGNA,RJ(90),RHO PUN00130
0008      COMMON/CRST/DELX(45),DELO(45),RHOIN,RATIO,EPST,THETA0, PUN00140
      IRSIG1,G1,X1(45),X2(45),X3(45),XR2(45),XR1(45),PR1, PUN00150
      ZPR2,P1,F1,RJ1(90),DOTT,PGRAD(45),DIAG(45), PUN00160
      3 PREV3,ADLX, NTCTR, NUMINI, NPHASE, NSATIS PUN00170
0009      COMMON/EXOPT/NEXP1,NEXP2,NEXP3,NEXP4,NEXP5,XEP1,XEP2 PUN00180
0010      T=60.0 PUN00190
0011      WRITE (7,10) EPST,RHO,THETA0,RATIO,T,M,N,M1 PUN00200
C      TMAX=15 SET TO 60. SECONDS PUN00210
0012      WRITE (7,20) (X(I),I=1,N) PUN00220
0013      NT1=3 PUN00230
C      SET RHO OPTION SO THIS VALUE OF RHO WILL BE USE FOR THE RESTART. PUN00240
0014      WRITE (7,30) NT1,NT2,NT3,NT4,NT5,NT6,NT7,NT8,NT9,NT10 PUN00250
0015      WRITE (7,20) XEP1, XEP2 PUN00260
0016      WRITE (7,30) NEXP1,NEXP2,NEXP3 PUN00270
0017      RETURN PUN00280
C                      PUN00290
0018      10  FORMAT (5E12.5,3I4) PUN00300
0019      20  FORMAT (6E12.5) PUN00310
0020      30  FORMAT (10I7) PUN00320
0021      END PUN00330

```

Figure 29.--Subroutine PUNCH.

```

FORTAN IV G LEVEL 21          REJECT          DATE = 80242          12/13/03

0001      SUBROUTINE REJECT                      REJ00040
C                      OCTOBER 1970              REJ00050
C                      C REJECT RETURNS THE STORED VALUES OF THE OBJECTIVE FUNCTION, THE
C                      C CONSTRAINT FUNCTIONS AND THE PENALTY FUNCTION TO THEIR NORMAL LOCATION REJ00060
C                      C                      REJ00070
0002      IMPLICIT REAL*8(A-H,O-Z)              REJ00080
0003      REAL*4 RHOIN,RATIO,EPST,THETA0          REJ00090
0004      COMMON/SHARE/X(45),DEL(45),A(45,45),N,M,MN,NP1,NM1 REJ00100
0005      COMMON /EQUAL/ H, H1, H2              REJ00110
0006      COMMON/VALUE/F,G,PO,RSIGNA,RJ(90),RHO REJ00120
0007      COMMON/CRST/DELX(45),DELO(45),RHOIN,RATIO,EPST,THETA0, REJ00130
      IRSIG1,G1,X1(45),X2(45),X3(45),XR2(45),XR1(45),PR1, REJ00140
      ZPR2,P1,F1,RJ1(90),DOTT,PGRAD(45),DIAG(45), REJ00150
      3 PREV3,ADLX, NTCTR, NUMINI, NPHASE, NSATIS REJ00160
0008      DO 10 I=1,N REJ00170
0009      X(I)=X1(I) REJ00180
0010      MMZ=M+NZ REJ00190
0011      DO 20 J=1,MMZ REJ00200
0012      RJ(J)=RJ1(J) REJ00210
0013      PO=P1 REJ00220
0014      RSIGNA=RSIG1 REJ00230
0015      G=G1 REJ00240
0016      F=F1 REJ00250
0017      H=H1 REJ00260
0018      RETURN REJ00270
0019      END REJ00280

```

Figure 30.--Subroutine REJECT.

FURTHAN IV G LEVEL 21

RHUCOM

DATE = 80242

12/39/21

```

0001      SUBROUTINE RHOCOM                                RH000040
C                                                    RH000050
C                                                    RH000060
C                                                    RH000070
C                                                    RH000080
C                                                    RH000090
C                                                    RH000100
0002      SUBROUTINE TO COMPUTE INITIAL RHO VALUE
0003      CONTROLLED BY CUL. 7 ON OPTION CARD
0004      IMPLICIT REAL*8(A-H,O-Z)
0005      REAL*4 RHOIN,RATIO,EPST,THETA0
0006      COMMON/SHARE/X(45),DEL(45),A(45,45),N,M,MN,NP1,NM1
0007      COMMON /OPTNS/ NT1,NT2,NT3,NT4,NT5,NT6,NT7,NT8,NT9,NT10
0008      COMMON/VALUE/F,G,PO,RSIGMA,RJ(90),RHO
0009      COMMON/CRST/DELX(45),DELX0(45),RHOIN,RATIO,EPST,THETA0,
0010      IRSIG1,G1,X1(45),X2(45),X3(45),XR2(45),XR1(45),PR1,
0011      2PR2,P1,F1,RJ1(90),DOTT,PGRAD(45),DIAG(45),
0012      3 PREV3,ADDELX, NTCTR, NUMINI, NPHASE, NSATIS
0013      GO TO (110,50,10,190), NT1
0014      10 RHO=RHOIN
0015      20 IF (RHO) 30,30,40
0016      30 RHO=1.
0017      40 RETURN
0018      50 NPAR1=1
0019      60 RHO=1.
0020      C 2 MEANS RHO WHICH MINIMIZES GRADIENT MAG.
0021      CALL GRAD (2)
0022      DO 70 I=1,N
0023      PGRAD(I)=DELX0(I)
0024      RHO=2.
0025      CALL GRAD (2)
0026      DO 80 I=1,N
0027      DELX0(I)=DELX0(I)-PGRAD(I)
0028      PGRAD(I)=PGRAD(I)-DELX0(I)
0029      GO TO (90,130), NPAR1
0030      90 DOT1=0.
0031      DOT2=0.
0032      DO 100 I=1,N
0033      DOT1=DOT1+DELX0(I)*PGRAD(I)
0034      DOT2=DOT2+DELX0(I)**2
0035      RHO=DABS(DOT1/DOT2)
0036      GO TO 20
0037      C 3 MEANS COMPUTE RHO SO AS TO MINIMIZE DEL P1/DDP/1,DEL P
0038      110 NPAR2=1
0039      120 NPAR1=2
0040      C USE DF AND DR SUBROUTINE
0041      GO TO 60
0042      130 RHO=1.
0043      C ASSUME SIGMA TERM IS CONSID. GRTER THAN F TERM
0044      CALL SECORD (2)
0045      DO 140 I=1,N
0046      DELX(I)=PGRAD(I)
0047      CALL INVERS (1)
0048      DO 150 I=1,N
0049      X1(I)=DELX(I)
0050      DELX(I)=DELX0(I)
0051      CALL SECORD (2)
0052      CALL INVERS (1)
0053      DO 160 I=1,N
0054      XR2(I)=DELX(I)
0055      DOT1=0.
0056      DOT2=0.
0057      DO 180 I=1,N
0058      DOT1=DOT1+PGRAD(I)*X1(I)
0059      DOT2=DOT2+DELX0(I)*XR2(I)
0060      RHO=DSORT(DABS(DOT1/DOT2))
0061      GO TO 20
0062      190 NPAR2=2
0063      C RHO MINIMIZES 2ND ORDER MOVE
0064      GO TO 120
0065      C USES INTERNAL SUB. TO COM /DDP/-1 DF AND /DDP/- DR
0066      200 DOT1=0.0
0067      DOT2=0.0
0068      DO 210 I=1,N
0069      DOT1=X1(I)**2+DOT1
0070      DOT2=X1(I)*XR2(I)+DOT2
0071      RHO=DABS(DOT1/DOT2)
0072      GO TO 20
0073      END

```

Figure 31.--Subroutine RHOCOM.

FORTRAN IV G LEVEL 21	SECUND	DATE = 80242	12/14/02
0001	SUBROUTINE SECUND(SECS)		SEC00050
0002	CALL CLOCK(1)		SEC00060
0003	SECS = 1		SEC00070
0004	SECS = SECS / 100.0		SEC00080
0005	RETURN		SEC00090
0006	END		SEC00100

Figure 32.--Subroutine SECUND.

## 6.25 SECØRD

Subroutine SECØRD evaluates the matrix of second partials of the W function. It calls on the user-supplied subroutine MATRIX to evaluate the upper triangle of the matrix of the second partials of the objective and the constraint functions. Subroutine SECØRD is listed in Figure 33.

## 6.26 SENS

This subroutine calculates the solution point and optimal value function sensitivities for each subproblem (if required), as well as the final problem, by implementing Steps 1-3 and 7-8 of Algorithm 2.1 of Section 2. The various choices for sensitivity calculations are given in options 3, 4, and 5 of the second option card (see Table 4). SENS is called by the program MAIN and the subroutine BØDY. It calls subroutine LMULT to calculate the Lagrange multiplier sensitivities. SENS is shown as Figure 34.

## 6.27 SET

Subroutine SET is called once by MAIN at the start of the attempt to solve a problem. It obtains and stores the value of the computer's clock. It obtains the value of the clock by calling a system library or user coded subroutine that queries the computer's clock and returns its value as a floating point number in seconds. Figure 35 lists subroutine SET.

```

FORTRAN IV G LEVEL 21          SECOND          DATE = 80242          12/46/17

0001          SUBROUTINE SECORD (IS)                                SEC00040
C                                                                    SEC00050
C          OCTOBER 1970                                            SEC00060
C                                                                    SEC00070
C SECORD EVALUATES THE MATRIX OF SECOND PARTIALS OF THE PENALTY SEC00080
C FUNCTION.                                                         SEC00090
C                                                                    SEC00100
C- (1) MEANS DONT COMPUTE GRAD. OUTER PRODUCT (IN SECORD)        SEC00110
0002          IMPLICIT REAL*8(A-M,D-Z)                             SEC00120
0003          REAL*4 RHOIN,RATIO,EPSI,THETA0                       SEC00130
0004          COMMON/SHARE/X(45),DEL(45),A(45,45),N,M,MN,NP1,NM1 SEC00140
0005          COMMON /EQUAL/ H, H1, MZ                             SEC00150
0006          COMMON /OPTNS/ NT1,NT2,NT3,NT4,NT5,NT6,NT7,NT8,NT9,NT10 SEC00160
0007          COMMON/VALUE/F,G,PO,RSIGMA,RJ(90),RHO                SEC00170
0008          COMMON/CRST/DELX(45),DELX0(45),RHOIN,RATIO,EPSI,THETA0, SEC00180
          1RSIG1,G1,X1(45),X2(45),X3(45),XR2(45),XR1(45),PR1, SEC00190
          2PR2,P1,F1,RJ1(90),DOTT,PGRAD(45),DIAG(45), SEC00200
          3 PREV3,ADELX, NTCTR, NUMINI, NPHASE, NSATIS SEC00210
          DO 10 I=1,N                                             SFC00220
          DO 10 J=1,N                                             SEC00230
          A(I,J)=0.                                               SEC00240
          GO TO (230,20), IS SEC00250
CGRAD. TERM NOT PREV. COMPUTED SEC00260
          DO 30 I=1,N                                             SEC00270
          DO 30 J=1,I                                             SEC00280
          A(I,J)=0.0 SEC00290
          30 CONTINUE SEC00300
          GO TO (40,60), NT2 SEC00310
          DO 50 I=1,N                                             SEC00320
          A(I,J)=RHO/X(I)**2 SEC00330
          50 CONTINUE SEC00340
          60 IF (M.LE.0) GO TO 130 SEC00350
          DO 120 IN=1,M SEC00360
          IF (RJ(IN)) 120,120,70 SEC00370
          70 CALL GRAD1 (IN) SEC00380
          TT = RHO/RJ(IN)**2 SEC00390
          DO 110 I=1,N SEC00400
          IF (DEL(I)) 80,110,80 SEC00410
          T=TT*DEL(I) SEC00420
          80 DO 100 J=1,I SEC00430
          IF (DEL(J)) 90,100,90 SEC00440
          A(I,J)=A(I,J)+T*DEL(J) SEC00450
          90 CONTINUE SEC00460
          100 CONTINUE SEC00470
          110 CONTINUE SEC00480
          120 CONTINUE SEC00490
C EQUALITY CONSTRAINTS SEC00500
          130 IF (MZ) 210,210,140 SEC00510
          140 GO TO (210,150,230), NPHASE SEC00520
          150 RQ=2./RHO SEC00530
          DO 200 JJ=1,MZ SEC00540
          IN=M+JJ SEC00550
          CALL GRAD1 (IN) SEC00560
          DO 190 I=1,N SEC00570
          IF (DEL(I)) 160,190,160 SEC00580
          160 T=RQ*DEL(I) SEC00590
          DO 180 J=1,I SEC00600
          IF (DEL(J)) 170,180,170 SEC00610
          A(I,J)=A(I,J)+T*DEL(J) SEC00620
          170 CONTINUE SEC00630
          180 CONTINUE SEC00640
          190 CONTINUE SEC00650
          200 CONTINUE SEC00660
          210 DO 220 I=1,N SEC00670
          DIAG(I)=A(I,I) SEC00680
          220 A(I,I)=0. SEC00690
C READY NOW FOR MATRIX OF 2ND PARTIALS OF RESTRAINTS SEC00700
          230 GO TO (240,510,520), NT10 SEC00710
          240 IF (M.LE.0) GO TO 340 SEC00720
          DO 330 IN=1,M SEC00730
          LORN=2 SEC00740
C CONSTRAINT ASSUMED NONLINEAR SEC00750
          CALL MATRIX (IN,LORN) SEC00760
          IF (LORN.LT.2) GO TO 330 SEC00770
          IF (RJ(IN) .GT. 0.0) GO TO 280 SEC00780
          DO 260 I=2,M SEC00790
          IM1=I-1 SEC00800
          DO 260 J=1,IM1 SEC00810
          IF (A(J,I)) 250,260,250 SEC00820
          250 A(I,J)=A(I,J)-A(J,I) SEC00830
          A(J,I)=0. SEC00840
          260 CONTINUE SEC00850
          DO 270 I=1,N SEC00860
          DIAG(I)=DIAG(I)-A(I,I) SEC00870
          270 A(I,I)=0.0 SEC00880
          GO TO 330 SEC00890
          280 T=-RHO/RJ(IN) SEC00900
          DO 300 I=2,M SEC00910
          IM1=I-1

```

Figure 33.--Subroutine SECORD.

0074	DO 300 J=1,IM1	SEC00910
0075	IF (A(J,I)) 290,300,290	SEC00920
0076	290 A(I,J)=A(I,J)+T*A(J,I)	SEC00930
0077	A(J,I)=0.	SEC00940
0078	300 CONTINUE	SEC00950
0079	DO 320 I=1,N	SEC00960
0080	IF (A(I,I)) 310,320,310	SEC00970
0081	310 DIAG(I)=DIAG(I)+T*A(I,I)	SEC00980
0082	A(I,I)=0.	SEC00990
0083	320 CONTINUE	SEC01000
0084	330 CONTINUE	SEC01010
0085	340 CONTINUE	SEC01020
0086	GO TO (520,350,520), NPHASE	SEC01030
0087	350 IF (MZ.EQ.0) GO TO 420	SEC01040
	C-- EQUALITY SECOND PARTIALS HERE	SEC01050
0088	IF (INTIO.GE.2) GO TO 420	SEC01060
0089	DO 410 I=1,MZ	SEC01070
0090	IN=M+I	SEC01080
0091	LORN=2	SEC01090
0092	CALL MATRIX (IN,LORN)	SEC01100
0093	IF (LORN.LT.2) GO TO 410	SEC01110
0094	T=2.*RJ(IN)/RMD	SEC01120
0095	DO 380 I=2,N	SEC01130
0096	IM1=I-1	SEC01140
0097	DO 370 J=1,IM1	SEC01150
0098	IF (A(J,I)) 360,370,360	SEC01160
0099	360 A(I,J)=A(I,J)+T*A(J,I)	SEC01170
0100	A(J,I)=0.0	SEC01180
0101	370 CONTINUE	SEC01190
0102	380 CONTINUE	SEC01200
0103	DO 400 I=1,N	SEC01210
0104	IF (A(I,I)) 390,400,390	SEC01220
0105	390 DIAG(I)=DIAG(I)+T*A(I,I)	SEC01230
0106	A(I,I)=0.0	SEC01240
0107	400 CONTINUE	SEC01250
0108	410 CONTINUE	SEC01260
	C GET MATRIX OF 2ND PARTIALS OF OBJECTIVE FUNCTION	SEC01270
0109	420 LLL=2	SEC01280
0110	CALL MATRIX (0,LLL)	SEC01290
0111	IF (LLL.LT.2) GO TO 490	SEC01300
0112	DO 440 I=2,N	SEC01310
0113	IM1=I-1	SEC01320
0114	DO 440 J=1,IM1	SEC01330
0115	IF (A(J,I)) 430,440,430	SEC01340
0116	430 A(I,J)=A(I,J)+A(J,I)	SEC01350
0117	440 A(J,I)=A(I,J)	SEC01360
0118	DO 470 I=1,N	SEC01370
0119	IF (A(I,I)) 450,460,450	SEC01380
0120	450 A(I,I)=DIAG(I)+A(I,I)	SEC01390
0121	GO TO 470	SEC01400
0122	460 A(I,I)=DIAG(I)	SEC01410
0123	470 CONTINUE	SEC01420
0124	480 RETURN	SEC01430
0125	490 DO 500 I=1,N	SEC01440
0126	A(I,I)=DIAG(I)	SEC01450
0127	DO 500 J=1,N	SEC01460
0128	A(I,J)=A(J,I)	SEC01470
0129	GO TO 480	SEC01480
0130	510 GO TO (520,350,350), NPHASE	SEC01490
0131	520 DO 530 I=2,N	SEC01500
0132	IM1=I-1	SEC01510
0133	DO 530 J=1,IM1	SEC01520
0134	A(J,I)=A(I,J)	SEC01530
0135	DO 540 I=1,N	SEC01540
0136	540 A(I,I)=DIAG(I)	SEC01550
0137	GO TO 480	SEC01560
0138	END	SEC01570

Figure 33.--continued

FURTRAN IV G LEVEL 21

SENS

DATE = 80242

12/27/55

```

0001      SUBROUTINE SENS                                SEN00060
C                                                    SEN00070
C              1 MARCH 1976                            SEN00080
C                                                    SEN00090
C THIS VERSION OF THE SENSITIVITY ANALYSIS SUBROUTINE IS USED TO    SEN00100
C COMPUTE THE DIRECTIONAL DERIVATIVES OF X AND F WITH RESPECT TO    SEN00110
C CERTAIN PARAMETERS CODED IN THE ARRAY PAR(20). THE DIRECTIONAL    SEN00120
C DERIVATIVES ARE ESTIMATED FOR ONE PARAMETER AT A TIME WITH NPAR    SEN00130
C THE NUMBER OF PARAMETERS INVOLVED IN THE SENSITIVITY ANALYSIS. THE SEN00140
C USE OF THE PARAMETERS PAR(20) MUST BE CONSISTENT THROUGHOUT THE    SEN00150
C USER'S SUBROUTINES.                                             SEN00160
C THE SUBROUTINE IS USED FOR A SENSITIVITY ANALYSIS AT THE FINAL SUB- SEN00170
C PROBLEM OR FOR A SENSITIVITY ANALYSIS AT EACH SUBPROBLEM ALONG THE SEN00180
C MINIMIZING TRAJECTORY. DPAR(20) IS THE ARRAY OF DIFFERENCING    SEN00190
C INTERVALS CORRESPONDING TO THE PARAMETERS PAR(20). DPAR(20) IS    SEN00200
C ASSIGNED VALUES IN SUBROUTINE PARDIF.                            SEN00210
C THIS APPROACH TO SENSITIVITY ANALYSIS IS DUE TO A. V. FIACCO. THE    SEN00220
C FIRST VERSION WAS CODED BY B. CAUSEY. THE SECOND VERSION WAS CODED    SEN00230
C BY M. C. MYLANDER. THE THIRD VERSION WAS AN EXTENSION OF THE    SEN00240
C SECOND VERSION TO PERMIT SENSITIVITY ANALYSIS ALONG THE MINIMIZING    SEN00250
C TRAJECTORY, AND WAS CODED BY R. L. ARMACOST. THIS IS THE FOURTH    SEN00260
C VERSION WHICH INCORPORATES THE OPTION TO CALCULATE THE PARTIAL    SEN00270
C DERIVATIVES OF THE LAGRANGE MULTIPLIERS (SUBROUTINE LMULT) AND TO    SEN00280
C CONDUCT A PRELIMINARY SCREENING OF THE PARAMETERS BY ESTIMATING THE SEN00290
C FIRST ORDER SENSITIVITY OF THE OPTIMAL VALUE FUNCTION (SUBROUTINE    SEN00300
C PRESEN). THIS VERSION WAS CODED BY R. L. ARMACOST.                SEN00310
C THIS ROUTINE TOGETHER WITH OTHER SENSITIVITY ROUTINE WERE         SEN00320
C REDIMENSIONED TO 45 I.E. X(45) AND PAR(45). FURTHERMORE         SEN00330
C IT WAS SLIGHTLY MODIFIED TO INTERFACE BOUND CALCULATION          SEN00340
C ROUTINES. GHAEMI(1979).                                          SEN00350
0002      IMPLICIT REAL*8(A-H,O-Z)                               SEN00360
0003      REAL*4 RHOIN,RATIO,EPSI,THETA0,XEP1,XEP2                 SEN00370
0004      COMMON/SHARE/X(45),DEL(45),A(45,45),N,M,MN,NP1,NM1      SEN00380
0005      COMMON/EQUAL/H,M1,M2                                     SEN00390
0006      COMMON/OPTNS/NT1,NT2,NT3,NT4,NT5,NT6,NT7,NT8,NT9,NT10    SEN00400
0007      COMMON/VALUE/F,G,PO,RSIGMA,RJ(90),RHO                  SEN00410
0008      COMMON/CRST/DELX(45),DELX0(45),RHOIN,RATIO,EPSI,THETA0, SEN00420
      INTCTR,NUMINI,X1(45),X2(45),X3(45),XR2(45),XR1(45),PR1,    SEN00430
      ZPR2,P1,F1,RJ1(90),DOTT,PGRAD(45),DIAG(45),                SEN00440
      3PREV3,ADDELX,RSIG1,G1,NPHASE,NSATIS                       SEN00450
      COMMON/SEN/PAR(45),DPAR(45),NPAR,ISENS                     SEN00460
      COMMON/EXPOPT/NEXOP1,NEXOP2,NEXOP3,NEXOP4,NEXOP5,XEP1,XEP2 SEN00470
      COMMON/ABG/LY,LZ,PER(45)                                    SEN00480
      COMMON/ABG1/FEL,FE2                                         SEN00490
      COMMON/ABG2/DF1(45),DF2(45)                                 SEN00500
      COMMON/ABG5/TX(45),TDELX(45)                                SEN00510
      COMMON/OP6/NEXOP6                                           SEN00520
      DIMENSION DELT(45),DELTX(45),X2H(45),X3H(45)              SEN00530
      DIMENSION DELMU(90),DUM(90),KTEST(45)                      SEN00540
      DO 5 I=1,N                                                  SEN00550
      DELTX(I) = DELX(I)                                          SEN00560
      5 DELT(I) = DELX0(I)                                         SEN00570
      CALL STORE                                                  SEN00580
      AVAL = 1.0E-10                                             SEN00590
      MPHZ = M + MZ                                              SEN00600
      ISENS = ISENS + 1                                           SEN00610
      CALL PARDIF                                                 SEN00620
C WRITE OUT X, RHO AND PAR.                                       SEN00630
0026      WRITE(6,10) RHO                                         SEN00640
0027      10 FORMAT(1H //30X,20HSENSITIVITY ANALYSIS //          SEN00650
      15X,22HTHE VALUE OF R(RHO) IS ,E12.5)                      SEN00660
      WRITE(6,20)                                                 SEN00670
0028      20 FORMAT(16X, 64HTHE POINT AT WHICH THE ESTIMATE OF SENSITIVITY WILL SEN00680
0029      1BE MADE IS )                                           SEN00690
      DO 40 I=1,N,6                                               SEN00700
      II=MNO(I+5,N)                                              SEN00710
      WRITE(6,30) ((J,X(J)), J=I,II)                             SEN00720
      30 FORMAT( 6(2X,2MX(,12,2H)=,G14.7) )                     SEN00730
      40 CONTINUE                                                 SEN00740
C CALCULATION OF GP VARIABLE VALUE                                SEN00750
      IF(NEXOP6.NE.1) GO TO 660                                  SEN00760
      INDEX=0.                                                    SEN00770
      CALL TRANS(INDEX)                                           SEN00780
      WRITE(6,400)                                                SEN00790
      400 FORMAT(16X,'CORRESPONDING POINT FOR GP PROBLEM IS')    SEN00800
      DO 640 I=1,N,6                                              SEN00810
      II=MNO(I+5,N)                                              SEN00820
      640 WRITE(6,30)((J,X(J)),J=I,II)                             SEN00830
      DO 650 I=1,N                                               SEN00840
      650 X(I)=TX(I)                                             SEN00850
      660 CONTINUE                                                SEN00860
      WRITE(6,50) ((I,PAR(I),DPAR(I)), I=1,NPAR)                SEN00870
      50 FORMAT(45H0      PARAMETER VALUE      DIFFERENCING INTERVAL / SEN00880
      1 (14,2X,G14.6,6X,G14.5) )                                SEN00890
      WRITE(6,55)                                                 SEN00900
      55 FORMAT(//)                                               SEN00910
      EVALUATE F, G AND H.                                         SEN00920

```

Figure 34.--Subroutine SENS.



```

0050      CALL RESTNT(I,F)
0051      IF(MPM2.EQ.0) GO TO 85
0052      DO 80 J=1,NPM2
0053      CALL RESTNT(I,RJ(I))
0054      80 CONTINUE
0055      IF(NEXP5.EQ.2) GO TO 85
0056      CALL PRESENIDU,KTEST)
0057      IF(LY.GT.0) GO TO 400

C COMPUTE DEL F.
0058      85 CALL GRAD1(I)
0059      DO 90 I=1,N
0060      X3H(I) = DEL(I)
0061      90 CONTINUE

C COMPUTE (DEL)**2 P - STORED IN A.
0062      CALL SECORD(2)

C PERFORM THE L-U DECOMPOSITION OF A.
0063      DO 100 I=1,N
0064      DELX(I)=0.0
0065      100 CONTINUE
0066      NP=NT3
0067      NT3=1
0068      CALL INVERS(1)

C CHECK TO MAKE SURE AN ORTHOGONAL MOVE IS NOT ATTEMPTED.
0069      DO 110 I=1,N
0070      IF(DELX(I).EQ.0.0) GO TO 110
0071      WRITE(6,105)
0072      105 FORMAT(94H0 THE MATRIX OF SECOND PARTIALS IS NOT POSITIVE DEFINITE
1, SENSITIVITY ANALYSIS IS TERMINATED )
GO TO 202

0073      110 CONTINUE
0074      DO 120 I=1,NPAR
0075      X2H(I) = PAR(I)
0076      120 CONTINUE
0077      DO 200 J=1,NPAR
0078      IF(NEXP5.NE.0) GO TO 115
0079      IF(KTEST(J).EQ.0) GO TO 200
0080      115 IF(NEXP4.EQ.0) GO TO 121
0081      CALL LMULT(0,DELMU,DEM,DU)
0082      C COMPUTE D(DEL P)/DA(I) AND D(F)/DA(I) USING CENTRAL DIFFERENCING.
0083      121 PAR(J) = PAR(J) + DPAR(J)
0084      CALL RESTNT(0,DF)
0085      IF(M.EQ.0) GO TO 126
0086      DO 125 I=1,M
0087      CALL RESTNT(I,RJ(I))
0088      IF(RJ(I).GT.AVAL) GO TO 125
0089      123 DPAR(J)=0.1*DPAR(J)
0090      PAR(J) = X2H(J)
0091      WRITE(6,124) J,DPAR(J)
0092      124 FORMAT(16H RESETTING DPAR(I,2,3H)= ,G14.5)
0093      IF(DPAR(J).EQ.0.1) GO TO 201
0094      GO TO 121
0095      125 CONTINUE
0096      126 IF(MZ.EQ.0) GO TO 128
0097      DO 127 I=1,MZ
0098      MZPI=M+1
0099      CALL RESTNT(MZPI,RJ(MZPI))
0100      127 CONTINUE
0101      128 IF(NEXP4.EQ.0) GO TO 129
0102      CALL LMULT(1,DELMU,DEM,DU)
0103      129 CALL GRAD(2)
0104      DO 130 I=1,N
0105      DELX(I)=DELX0(I)
0106      130 CONTINUE
0107      DEM=2.0*DPAR(J)
0108      PAR(J)=PAR(J) - DEM
0109      CALL RESTNT(0,VAL)
0110      IF(M.EQ.0) GO TO 136
0111      DO 135 I=1,M
0112      CALL RESTNT(I,RJ(I))
0113      IF(RJ(I).GT.AVAL) GO TO 135
0114      GO TO 123
0115      135 CONTINUE
0116      136 IF(MZ.EQ.0) GO TO 138
0117      DO 137 I=1,MZ
0118      MZPI=M+1
0119      CALL RESTNT(MZPI,RJ(MZPI))
0120      137 CONTINUE
0121      138 IF(NEXP4.EQ.0) GO TO 139
0122      CALL LMULT(2,DELMU,DEM,DU)
0123      139 CALL GRAD(2)
0124      DF=(DF-VAL)/DEM
0125      DO 140 I=1,N
0126      DELX(I)=(DELX(I) - DELX0(I))/DEM
0127      140 CONTINUE
0128      PAR(J) = X2H(J)

C HAVING ALREADY FACTORED A, SOLVE A*X=B FOR X,
C WHERE B=DELX AND X=DX/DA(I).

```

Figure 34.--continued

```

0129      CALL INVERS(2)
C PRINT OUT DX/DA(IJ)
0130      WRITE(6,150) J
0131      150 FORMAT(47H X-DERIVATIVES ARE WITH RESPECT TO PARAMETER ,12)
0132      DO 170 I=1,N,6
0133      II=MIND(I+5,N)
0134      WRITE(6,160) (JJ,DELX(IJ)), JJ=1,II)
0135      160 FORMAT(614H DX(,12,2H)=,G14.7) )
0136      170 CONTINUE
C CALCULATION OF GP VARIABLE SENSITIVITY
0137      IF(NEXOP6.NE.1) GO TO 670
0138      INDEX=1
0139      CALL TRANS(INDEX)
0140      WRITE(6,610) J
0141      610 FORMAT(2X,'CORRESPONDING DERIVATIVES OF GP VARIABLES WRT'
0142      I,' PARAMETER',12,'ARE')
0143      DO 680 I=1,N,6
0144      II=MIND(I+5,N)
0145      WRITE(6,160) (JJ,DELX(IJ)), JJ=1,II)
0146      680 CONTINUE
0147      DO 690 I=1,N
0148      690 DELX(I)=TDELX(I)
0149      670 CONTINUE
0150      IF(NEXOP4.EQ.0) GO TO 375
0151      IF(N.EQ.0) GO TO 301
0152      CALL LMULT(3,DELMU,DEM,DU)
0153      WRITE(6,350) J
0154      350 FORMAT(/ 41H U-DERIVATIVES WITH RESPECT TO PARAMETER ,12)
0155      DO 351 I=1,M,6
0156      II = MIND(I+5,M)
0157      WRITE(6,352) ((JJ,DU(IJ)),JJ=1,II)
0158      352 FORMAT(614H DU(,12,2H)=,G14.7) )
0159      351 CONTINUE
0160      301 IF(MZ.EQ.0) GO TO 375
0161      CALL LMULT(4,DELMU,DEM,DU)
0162      WRITE(6,360) J
0163      360 FORMAT(/ 41H W-DERIVATIVES WITH RESPECT TO PARAMETER ,12)
0164      DO 361 I=1,MZ,6
0165      II = MIND(I+5,MZ)
0166      WRITE(6,362) ((JJ,DU(IJ+M)), JJ=1,II)
0167      362 FORMAT(614H DW(,12,2H)=,G14.7) )
0168      361 CONTINUE
0169      375 CONTINUE
C COMPUTE DF/DA(IJ).
0170      DO 180 I=1,N
0171      DF = DF + X3H(I)*DELX(I)
0172      180 CONTINUE
C PRINT DF/DA(IJ).
0173      WRITE(6,190) DF
0174      190 FORMAT(/ 10X,13HDF(X(H))/DA= ,G14.6/10H *****/)
0175      200 CONTINUE
0176      GO TO 202
0177      106 FORMAT(1H ,21INTERMINATING PARAMETER,13,16H DUE TO DPAR = 0 /)
0178      GO TO 200
0179      202 NTJ=NP
0180      CALL REJECT
0181      DO 205 I=1,N
0182      DELX(I) = DELTX(I)
0183      205 DELXO(I) = DELT(I)
0184      400 CONTINUE
0185      500 RETURN
0186      END

```

```

SEN01800
SEN01810
SEN01820
SEN01830
SEN01840
SEN01850
SEN01860
SEN01870
SEN01880
SEN01890
SEN01900
SEN01910
SEN01920
SEN01930
SEN01940
SEN01950
SEN01960
SEN01970
SEN01980
SEN01990
SEN02000
SEN02010
SEN02020
SEN02030
SEN02040
SEN02050
SEN02060
SEN02070
SEN02080
SEN02090
SEN02100
SEN02110
SEN02120
SEN02130
SEN02140
SEN02150
SEN02160
SEN02170
SEN02180
SEN02190
SEN02200
SEN02210
SEN02220
SEN02230
SEN02240
SEN02250
SEN02260
SEN02270
SEN02280
SEN02290
SEN02300
SEN02310
SEN02320
SEN02330
SEN02340
SEN02350
SEN02360
SEN02370
SEN02380
SEN02390
SEN02400
SEN02410
SEN02420

```

Figure 34.--continued

```

FORTRAN IV G LEVEL 21          SET          DATE = 80242          12/14/52

0001          SUBROUTINE SET (TMAX)          SET00040
          C          SET00050
          C          FEBRUARY 1971          SET00060
          C          SET00070
          C SET STORES THE TIME AT WHICH THE PROBLEM IS BEGUN          SET00080
          COMMON /TSM/ NSMW          SET00090
0002          COMMON /TMX/ TMO,EXT,EXT90          SET00100
0003          C          SET00110
          C          SET00120
          C SECOND GIVES JOB CPU EXECUTION TIME IN 1/1000 OF A SECOND          SET00130
          C          SET00140
          C          SET00150
          C          SET00160
          C          SET00170
          C          SET00180
          C          SET00190

0004          CALL SECOND (TMO)
0005          EXT=TMAX*TMO
0006          EXT90= TMO + 0.90*TMAX
0007          NSMW=1
0008          RETURN
0009          END

```

Figure 35.--Subroutine SET.

## 6.28 STØRE

Subroutine STØRE stores the values of the current point  $x$  and the associated values of  $F$ ,  $W$ , and  $G$  in a temporary storage area. These values are restored by a call to REJECT. The listing for subroutine STØRE appears as Figure 36.

```

FORTRAN IV G LEVEL 21          STORE          DATE = 80242          12/15/22

0001          SUBROUTINE STORE          ST000040
          C          ST000050
          C          OCTOBER 1970          ST000060
          C          ST000070
          C STORE STORES THE VALUES OF THE CURRENT POINT AND THE ASSOCIATED          ST000080
          C VALUES OF THE FUNCTIONS IN A TEMPORARY AREA.          ST000090
          C          ST000100
          C          ST000110
          C          ST000120
          C          ST000130
          C          ST000140
          C          ST000150
          C          ST000160
          C          ST000170
          C          ST000180
          C          ST000190
          C          ST000200
          C          ST000210
          C          ST000220
          C          ST000230
          C          ST000240
          C          ST000250
          C          ST000260
          C          ST000270
          C          ST000280
          C          ST000290
          C          ST000300

0002          IMPLICIT REAL*8(A-H,O-Z)
0003          REAL*4 RMOIN,RATIO,EPSI,THETA0
0004          COMMON/SHARE/X(45),DEL(45),A(45,45),N,M,MN,NP1,NM1
0005          COMMON /EQUAL/ M, M1, M2
0006          COMMON/VALUE/F,G,P0,RSIGMA,RJ(90),RMD
0007          COMMON/CRST/DELX(45),DELX0(45),RMOIN,RATIO,EPSI,THETA0,
          1RSIG1,G1,X1(45),X2(45),X3(45),XR2(45),XRI(45),PRI,
          2PRZ,P1,F1,RJ1(90),DOTT,PGRAD(45),DIAG(45),
          3PREV3,ADELX, NTCTR, NUMINI, NPHASE, NSATIS

0008          DO 10 I=1,N
0009          X1(I)=X(I)
0010          MNZ=M+M2
0011          DO 20 J=1,MNZ
0012          RJ1(J)=RJ(J)
0013          P1=P0
0014          F1=F
0015          G1=G
0016          RSIG1=RSIGMA
0017          M1=M
0018          RETURN
0019          END

```

Figure 36.--Subroutine STØRE.

## 6.29 TCHECK

If the print option has been set so that printouts occur only after a subproblem has been called, then TCHECK is called after each iteration of the algorithm used to minimize the W function. The elapsed time is computed as is done in TIMEC but it is not printed out. If the elapsed time exceeds 90 percent of the estimated time limit, then the print option is changed so a printout occurs after each iteration of the procedure being used to minimize the W function. If the elapsed time exceeds the user specified time limit, then the routine will cause termination of the attempt to solve the problem by setting NSW equal to 2. Subroutine TCHECK appears as Figure 37.

```

FORTRAN IV G LEVEL 21          TCHECK          DATE = 80243          10/40/07

0001          SUBROUTINE TCHECK                      TCM00050
              C          FEBRUARY 1971                TCM00060
              C          TCM00070
              C TCHECK CHECKS THE NUMBER OF SECONDS THAT HAVE ELAPSED SINCE THE START TCM00080
              C OF THE PROBLEM. IF THE SOLUTION IS TAKING LONGER THAN 90 PER-CENT TCM00090
              C OF THE ESTIMATED MAXIMUM TIME, A SWITCH IS SET TO GIVE MORE OUTPUT. TCM00100
0002          COMMON /TSW/ NSW                        TCM00110
0003          COMMON /OPTNS/ NT1,NT2,NT3,NT4,NT5,NT6,NT7,NT8,NT9,NT10 TCM00120
0004          COMMON /TMX/ TMO,EXT,EXT90              TCM00130
0005          CALL SECUND (SECS)                      TCM00140
0006          IF (SECS.LT.EXT90) RETURN                TCM00150
              C GETTING CLOSE TO EXCEEDING THE TIME LIMIT SET OUTPUT OPTION TO GIVE TCM00160
              C MORE OUTPUT.                          TCM00170
0007          NT3=2                                    TCM00180
0008          X=SECS - TMO                             TCM00190
0009          WRITE(6,10) X                            TCM00200
0010          10  FORMAT (6X,5HTIME=,F9.3,8H SECONDS ) TCM00210
0011          IF (SECS .GT. EXT) NSW=2                TCM00220
0012          CALL OUTPUT (1)                          TCM00230
0013          RETURN                                    TCM00240
0014          END                                      TCM00250

```

Figure 37.--Subroutine TCHECK.

## 6.30 TIMEC

Subroutine TIMEC calls the same routine as called by SET to obtain the current status of the computer's time clock. It then prints out the elapsed time since the call to SET. If the elapsed time is greater than the user specified maximum time limit, TIMEC will cause the termination of the attempt to solve the problem by setting NSW

equal to 2. The listing for subroutine TIMEC is shown below in Figure 38.

FORTRAN IV G LEVEL 21		TIMEC	DATE = 80242	12/15/53
0001		SUBROUTINE TIMEC		TIM00030
	C			TIM00040
	C	FEBRUARY 1971		TIM00050
	C			TIM00060
	C	TIMEC CHECKS THE NUMBER OF SECONDS THAT HAVE ELAPSED SINCE THE START		TIM00070
	C	OF THE PROBLEM. IT PRINTS THIS NUMBER. IF THE SOLUTION IS TAKING		TIM00080
	C	LONGER THAN THE ESTIMATED MAXIMUM TIME, A SWITCH IS SET TO TERMINATE		TIM00090
	C	THE RUN.		TIM00100
0002		COMMON /TSW/ NSWM		TIM00110
0003		COMMON /TMX/ TMQ,EXT,EXT90		TIM00120
	C			TIM00130
	C	SECOND GIVES JOB CPU EXECUTION TIME IN 1/1000 OF A SECOND		TIM00140
	C			TIM00150
0004		CALL SECOND (SECS)		TIM00160
0005		X=SECS-TMO		TIM00170
0006		WRITE (6,20) X		TIM00180
0007		IF (SECS.LT.EXT) GO TO 10		TIM00190
0008		NSWM=2		TIM00200
0009	10	RETURN		TIM00210
	C			TIM00220
0010	20	FORMAT (6X,5HTIME=F9.3,0H SECONDS)		TIM00230
0011		END		TIM00240

Figure 38.--Subroutine TIMEC.

### 6.31 TRANS

TRANS is a special purpose subroutine that performs an exponential transformation of the solution vector and sensitivity results. This subroutine is useful when solving problem  $c(x)$ , the convex equivalent of the geometric programming problem  $G(t)$ . The former problem is obtained from the latter via the transformation  $t = e^{-x_i}$ ,  $i=1, \dots, n$ . TRANS is used to transform the results to the original space of the variable  $t$ . The motivation for coding this subroutine was the computational difficulty experienced by some users while solving geometric programming problems with SUMT. TRANS is called by the subroutines OUTPUT and SENS. Figure 39 is a listing of subroutine TRANS.

### 6.32 XMØVE

Subroutine XMØVE contains most of the logic of the algorithm used to minimize the  $W$  function for a given value of  $r$  (RHØ). First, a

PONTAN IV G LEVEL 21	TRANS	DATE = 80243	12/19/30
0001	SUBROUTINE TRANS(INDEX)		TRA00050
	C SUBROUTINE TRANS CODED BY GHAEMI(1979) TRANSFORMS THE OPTIMAL		TRA00060
	CA* AND SENSITIVITY RESULTS TO T SPACE WHERE		TRA00070
	C T=EXP(-X). THIS ROUTINE IS USEFUL WHEN USER SOLVES THE CONVEX		TRA00080
	C EQUIVALENT OF GP PROBLEMS OBTAINED BY THE ABOVE EXPONENTIAL		TRA00090
	C TRANSFORMATION.		TRA00100
0002	IMPLICIT REAL*8(A-H,O-Z)		TRA00110
0003	COMMON/SHAPE/X(45),DEL(45),A(45,45),N,M,MN,NP1,NM1		TRA00120
0004	COMMON/CRST/DELX(45),DELX0(45),RHOIN,KATIO,EPSI,THETA0,		TRA00130
	IKS1,G1,X1(45),X2(45),X3(45),XR2(45),XR1(45),PR1,		TRA00140
	2PR2,P1,F1,RJ1(90),DOTT,PGRAD(45),DIAG(45),		TRA00150
	3 PHEV3,ADELX, NTCTR, NUMINI, NPHASE, NSATIS		TRA00160
0005	COMMON/ABG5/TX(45),TDELX(45)		TRA00170
0006	COMMON/ABG6/XX(45)		TRA00180
0007	IF(INDEX.EQ.1) GO TO 20		TRA00190
0008	DO 10 I=1,N		TRA00200
0009	TX(I)=X(I)		TRA00210
0010	X(I)=DEXP(-X(I))		TRA00220
0011	10 XX(I)=X(I)		TRA00230
0012	RETURN		TRA00240
0013	20 DO 30 I=1,N		TRA00250
0014	TDELX(I)=DELX(I)		TRA00260
0015	30 DELX(I)=-DEXP(-X(I))*DELX(I)		TRA00270
0016	RETURN		TRA00280
0017	END		TRA00290

Figure 39.--Subroutine TRANS.

direction is chosen in which it is believed the W function will initially decrease. Then OPT is used to find a minimum of the W function in that direction. Having generated a new point giving a lower value of the W function, XMØVE returns control to BODY.

Subroutine XMØVE contains three procedures for generating a direction vector. The choice of the procedure used is controlled by experimental option 2 (NEXØP2). If NEXØP2 = 1, then Newton's method is used to choose the direction vector S, that  $S^1 = -[V^2W(x^1, r_k)]^{-1} \nabla W(x^1)$ . If  $V^2W(x^1, r^k)$  is not a positive definite matrix,  $S^1$  is generated by rules outlined on page 167 in Fiacco and McCormick [10].

If NEXØP2 = 2, then the negative of the gradient is used as the direction vector,  $S^1 = -\nabla W(x^1, r^k)$ .

When NEXØP2 = 3 a variable matrix method is used to generate  $S^1$ . This method is McCormick's modification of the Fletcher-Power-Davidon, as described on pages 170-175 in Fiacco and McCormick [10]. Subroutine XMØVE is shown as Figure 40.

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```

FORTRAN IV G LEVEL 21      XMOVE      DATE = 80242      12/16/26

0001      SUBROUTINE XMOVE                                XMO00040
C                                                    XMO00050
C      MARCH 1971                                        XMO00060
C                                                    XMO00070
C XMOVE DETERMINES THE VECTOR ALONG WHICH THE SEARCH FOR A MINIMUM IS
C USING OPT.                                            XMO00080
0002      IMPLICIT REAL*8(A-H,O-Z)                    XMO00100
0003      REAL*4 RHOIN,RATIO,EPSI,THETA0,XEP1,XEP2      XMO00110
0004      COMMON/SHARE/X(45),DEL(45),A(45,45),N,M,MN,NP1,NM1 XMO00120
0005      COMMON/CRST/DELX(45),DELXO(45),RHOIN,RATIO,EPSI,THETA0,
1RSIG1,G1,X1(45),X2(45),X3(45),XR2(45),XR1(45),PR1,
2PR2,P1,F1,RJ1(90),DOTT,PGRAD(45),DIAG(45),
3PREV3,ADLX,NTCTR,NUMINI,NPHASE,NSATIS
0006      COMMON/EXOPT/NEXP1,NEXP2,NEXP3,NEXP4,NEXP5,XFP1,XEP2 XMO00170
0007      COMMON/XVE/SIG(45),YY(45),XXX(45),DELL(45)    XMO00180
C--NEXP2 DETERMINES HOW MOVE IS TO BE MADE           XMO00190
C NEXP2 = 1 USE MODIFIED NEWTON RAPHSON METHOD.        XMO00200
C           = 2 USE MODIFIED NEWTON RAPHSON METHOD, BUT ADD DELXO TO
C           ORTHOGONAL MOVE VECTOR IF HESSIAN IS INDEFINITE. XMO00210
C           = 3 USE STEEPEST DESCENT METHOD.            XMO00220
C           = 4 USE MCCORMICK'S MODIFICATION OF THE FLETCHER-POWELL
C           METHOD.                                     XMO00230
0008      GO TO (10,10,180,30), NEXP2                  XMO00240
C--NEWTON -RAPH WITH WHATEVER METHOD IS IN INVERSE    XMO00250
0009      10 CALL GRAD (1)                               XMO00260
C--ONE (1) MEANS ACCUMULATE MATRIX OF SECOND PARTIAL DERIVATIVES XMO00270
0010      CALL SECORD (1)                                XMO00280
0011      DO 20 I=1,N                                    XMO00290
0012      20 DELX(I)=DELXO(I)                            XMO00300
0013      CALL INVERS (1)                                XMO00310
C IF A NONPOSITIVE PIVOT IS ENCOUNTERED IN INVERSE AN ATTEMPT IS MADE TO XMO00320
C COMPUTE A VECTOR HAVING A POSITIVE DOT PRODUCT WITH A NEGATIVE
C EIGENVECTOR AND THE NEGATIVE OF DEL P.              XMO00330
0014      CALL STORE                                    XMO00340
0015      CALL OPT                                       XMO00350
0016      RETURN                                         XMO00360
C--F-P-D-MCC MOVE                                    XMO00370
0017      30 CALL GRAD (2)                               XMO00380
C--MN IS NO. OF MOVES FOR THIS VALUE OF RHO         XMO00390
0018      IF (MN.NE.0) GO TO 70                          XMO00400
0019      40 IREP=0                                       XMO00410
0020      IT=0                                            XMO00420
C--SET INITIAL GUESS INVERS MATRIX OF SECOND PARTIAL DERIVATIVES XMO00430
C-- USE PARTIAL INVERSE IF KNOWN                     XMO00440
0021      DO 50 I=1,N                                    XMO00450
0022      DO 50 J=1,N                                    XMO00460
0023      50 A(I,J)=0.0                                  XMO00470
0024      DO 60 I=1,N                                    XMO00480
0025      60 A(I,I)=1.0                                  XMO00490
0026      DO 80 I=1,N                                    XMO00500
0027      80 DELX(I)=DELXO(I)                            XMO00510
0028      IF (IREP.GT.N) GO TO 40                       XMO00520
0029      IF (IT.EQ.0) GO TO 130                        XMO00530
0030      DO 90 I=1,N                                    XMO00540
0031      SIG(I)=X(I)-XXX(I)                             XMO00550
0032      90 YY(I)=DELL(I)-DELXO(I)                     XMO00560
C--NEGATIVE GRADIENT STORED AND COMPUTED             XMO00570
C--COMPUTE MY                                         XMO00580
0033      DO 100 I=1,N                                   XMO00590
0034      DELX(I)=0.0                                    XMO00600
0035      DO 100 J=1,N                                   XMO00610
0036      100 DELX(I)=DELX(I)+A(I,J)*YY(J)               XMO00620
C--COMPUTE Y(SIG-MY)-1                               XMO00630
0037      ZCON=0.0                                       XMO00640
0038      DO 110 I=1,N                                   XMO00650
0039      110 ZCON=ZCON+YY(I)*(SIG(I)-DELX(I))           XMO00660
0040      IF (ZCON.EQ.0.0) GO TO 130                     XMO00670
0041      IREP=IREP+1                                    XMO00680
0042      ZC=1./ZCON                                     XMO00690
C-- UPDATE H MATRIX USING MCC FORMULA WHEN SCALAR NE 0 XMO00700
0043      DO 120 I=1,N                                   XMO00710
0044      T1=ZC*(SIG(I)-DELX(I))                         XMO00720
0045      DO 120 J=1,N                                   XMO00730
0046      A(I,J)=A(I,J)+T1*(DELX(J)+SIG(J))             XMO00740
0047      120 A(J,I)=A(I,J)                              XMO00750
C-- STORE CURRENT POINT AND CURRENT GRADIENT (NEG)   XMO00760
0048      130 DO 140 I=1,N                               XMO00770
0049      XXX(I)=X(I)                                     XMO00780
0050      DELL(I)=DELXO(I)                               XMO00790
0051      DO 150 I=1,N                                   XMO00800
0052      DELX(I)=0.0                                    XMO00810
0053      DO 150 J=1,N                                   XMO00820
0054      150 DELX(I)=DELX(I)+A(I,J)*DELXO(J)           XMO00830
0055      ZC1=0.0                                         XMO00840
0056      DO 160 I=1,N                                   XMO00850
0057      160 ZC1=DELX(I)**2+ZC1                         XMO00860
0058      ZC1=DSORT(ZC1)                                 XMO00870
XMO00880
XMO00890
XMO00900

```

Figure 40.--Subroutine XMOVE.

0059		DO 170 I=1,N	XMO00910
0060	170	DELX(I)=DELX(I)/ZC1	XMO00920
0061		CALL STORE	XMO00930
0062		CALL OPT	XMO00940
0063		IT=IT+1	XMO00950
0064		RETURN	XMO00960
0065	180	CONTINUE	XMO00970
		C STEEPEST DESCENT	XMO00980
0066		CALL GRAD (2)	XMO00990
0067		DO 190 I=1,N	XMO01000
0068	190	DELX(I)=DELX(I)	XMO01010
0069		CALL STORE	XMO01020
0070		CALL OPT	XMO01030
0071		RETURN	XMO01040
0072		END	XMO01050

Figure 40.--continued



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